

Image Processing

Unit-2

Section 2.3 to 2.6 in the text book*

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**Digital Image Processing by Rafael Gonzalez and Richard Woods*

Topics in this unit

- Image Sensing and Acquisition
- Image Sampling and Quantization
- Some Basic Relationships
- Nonlinear Operations

Image Sensing and Acquisition

- Image Acquisition Using a Single Sensor
- Image Acquisition Using Sensor Strips
- Image Acquisition Using Sensor Arrays
- A Simple Image Formation Model

Image Sampling and Quantization

- Basic Concepts in Sampling and Quantization
- Representing Digital Images
- Spatial and Gray-Level Resolution
- Aliasing and Moiré Patterns
- Zooming and Shrinking Digital Images

Basic Relationships Between Pixels

- Neighbors of a Pixel
- Adjacency, Connectivity, Regions, and Boundaries
- Distance Measures
- Image Operations on a Pixel Basis
- Linear and Nonlinear Operations

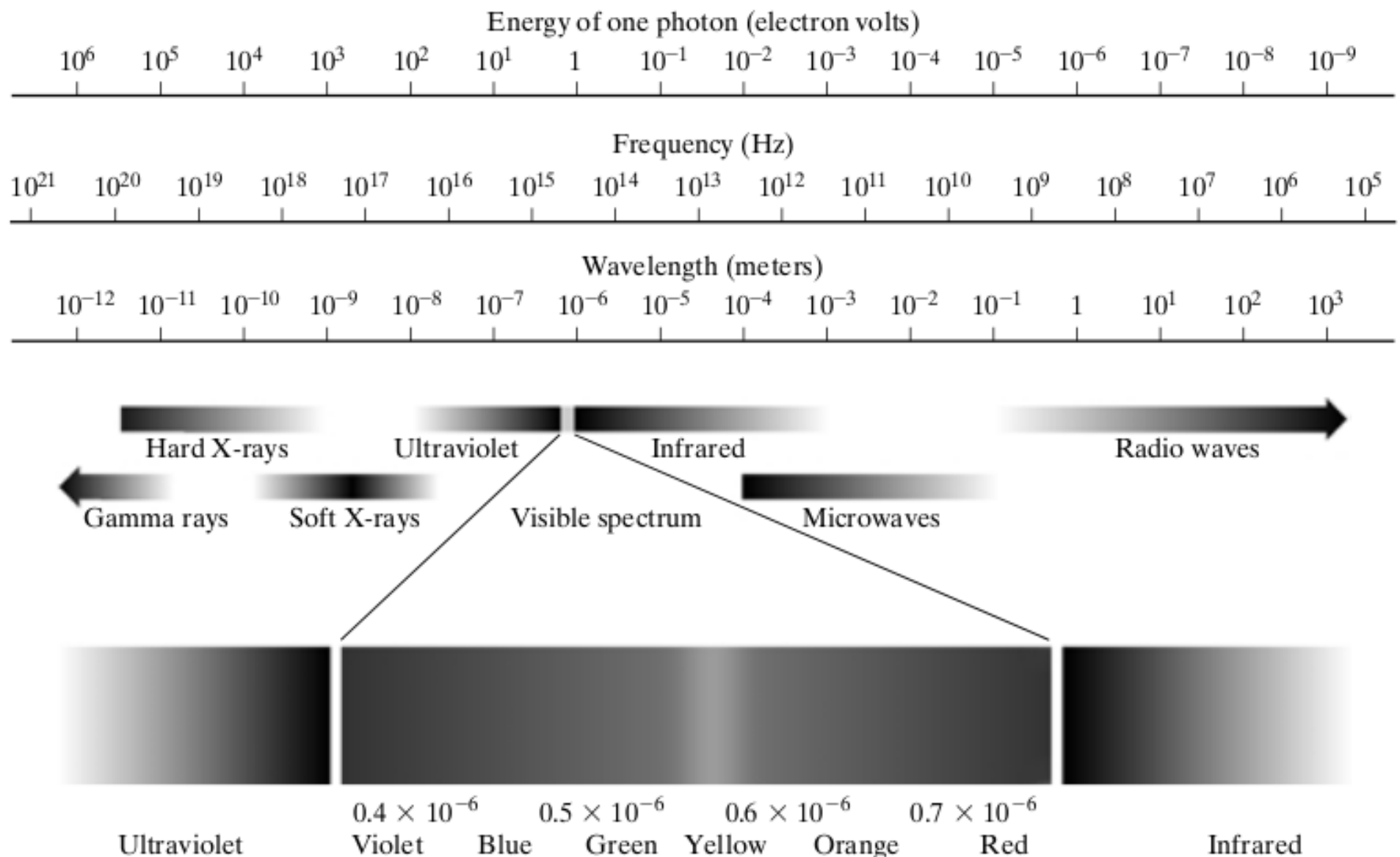


FIGURE 2.10 The electromagnetic spectrum. The visible spectrum is shown zoomed to facilitate explanation, but note that the visible spectrum is a rather narrow portion of the EM spectrum.

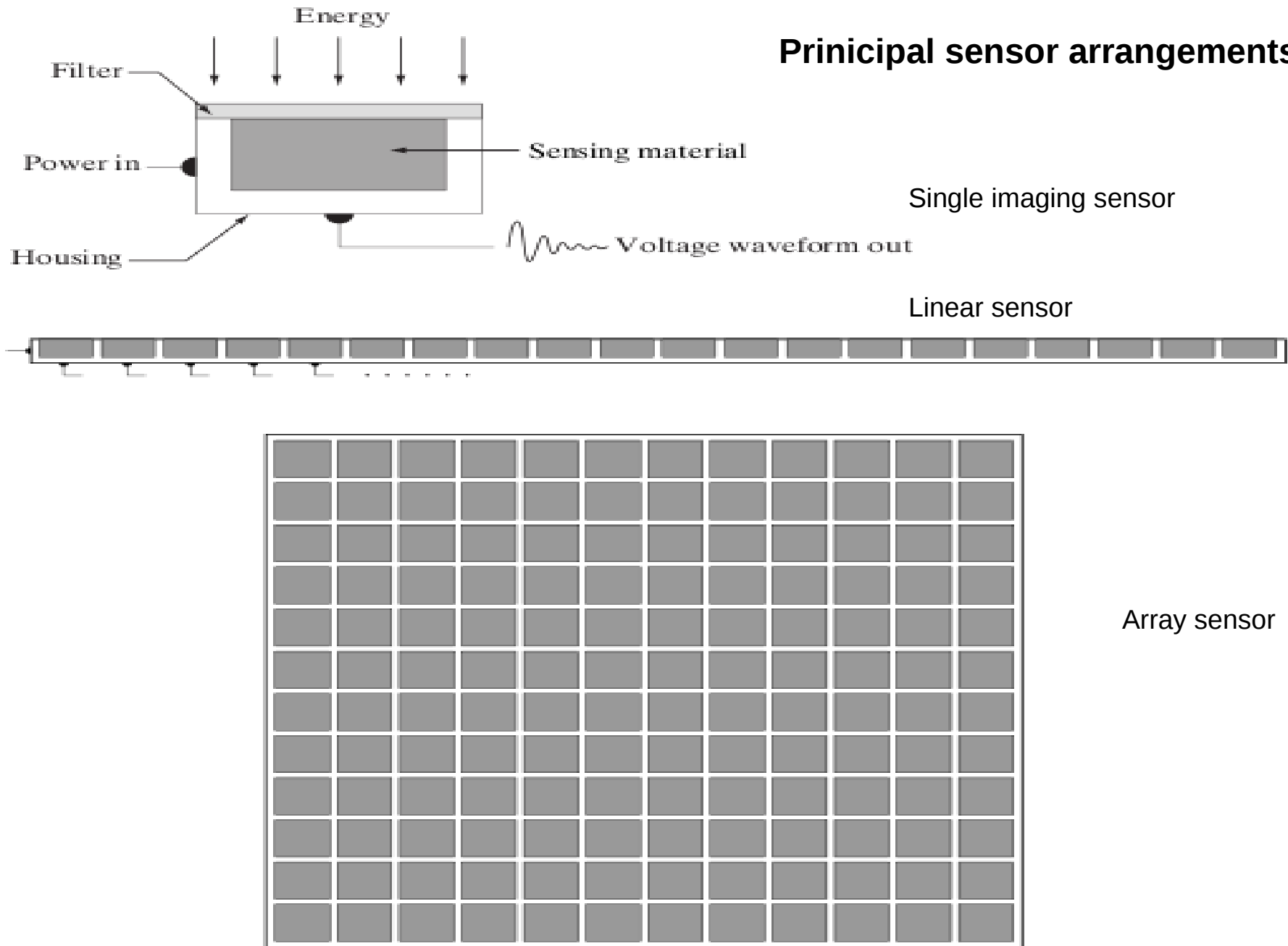
Image Sensing and Acquisition

Images are generated by the combination of an “illumination” source and the reflection or absorption of energy from that source by the elements of the “scene” being imaged.

Illumination : EM waves, ultrasound, computer generated image pattern

Scene: Molecules, buried rock formations, human brain ...

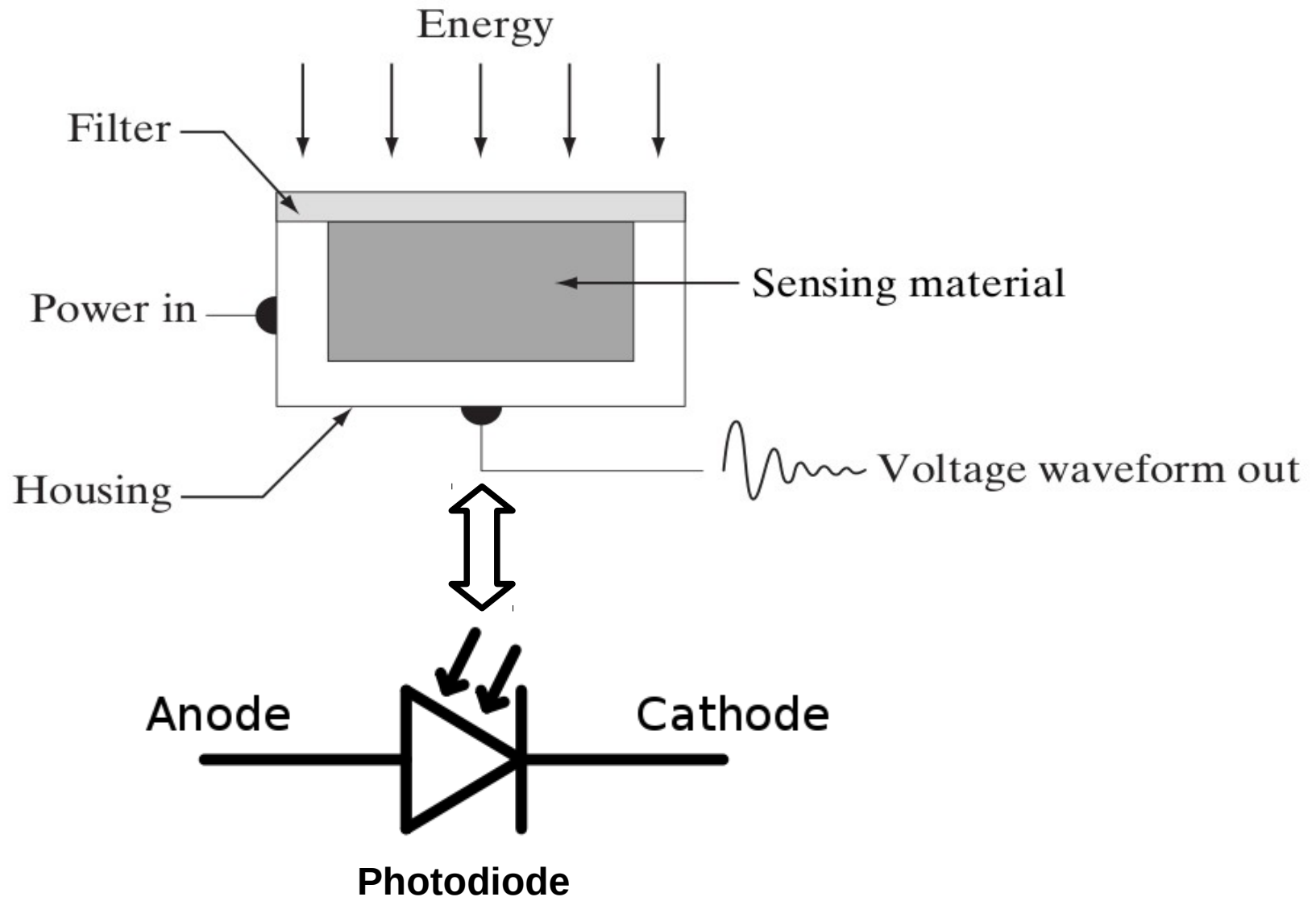
Principal sensor arrangements



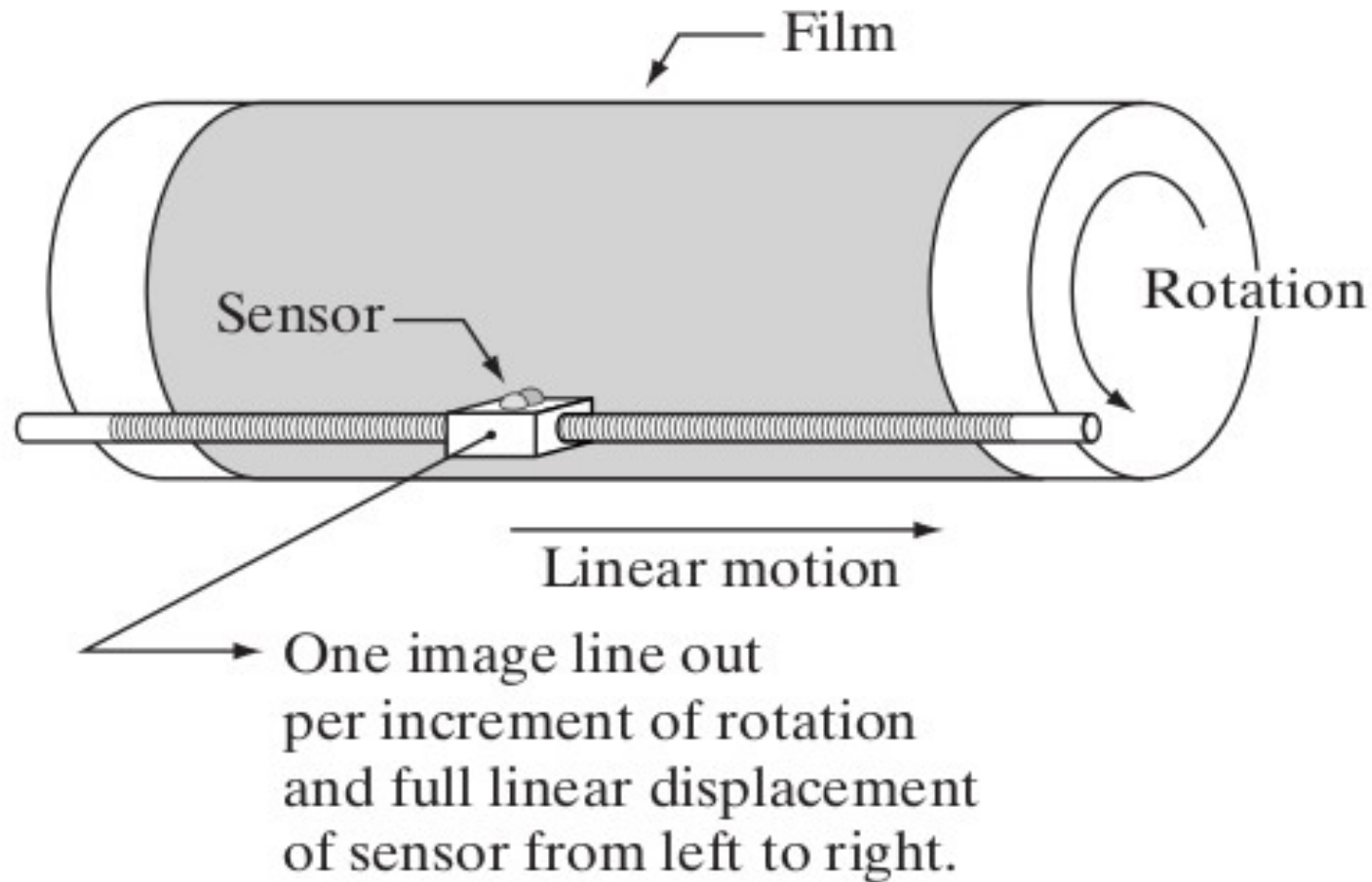
Principle behind image sensing

- Incoming energy is transformed into a voltage
- Combination of *input electrical power* and *sensor material* that is responsive to the particular type of energy being detected
- The output voltage waveform is the response of the sensor(s)
- A digital quantity is obtained from each sensor by digitizing its response

Image Acquisition Using a Single Sensor



2-D image using a single sensor



2-D image using a single sensor

- There has to be relative displacements in both the x- and y-directions between the sensor and the area to be imaged.
- A film negative is mounted onto a drum whose mechanical rotation provides displacement in one dimension
- The single sensor is mounted on a lead screw that provides motion in the perpendicular direction.
- Used in high-precision scanning
- **Adv:** Mechanical motion can be controlled with high precision, this method is an inexpensive
- **Disadv:** It is a slow method to obtain high-res images

Image Acquisition Using Sensor Strips

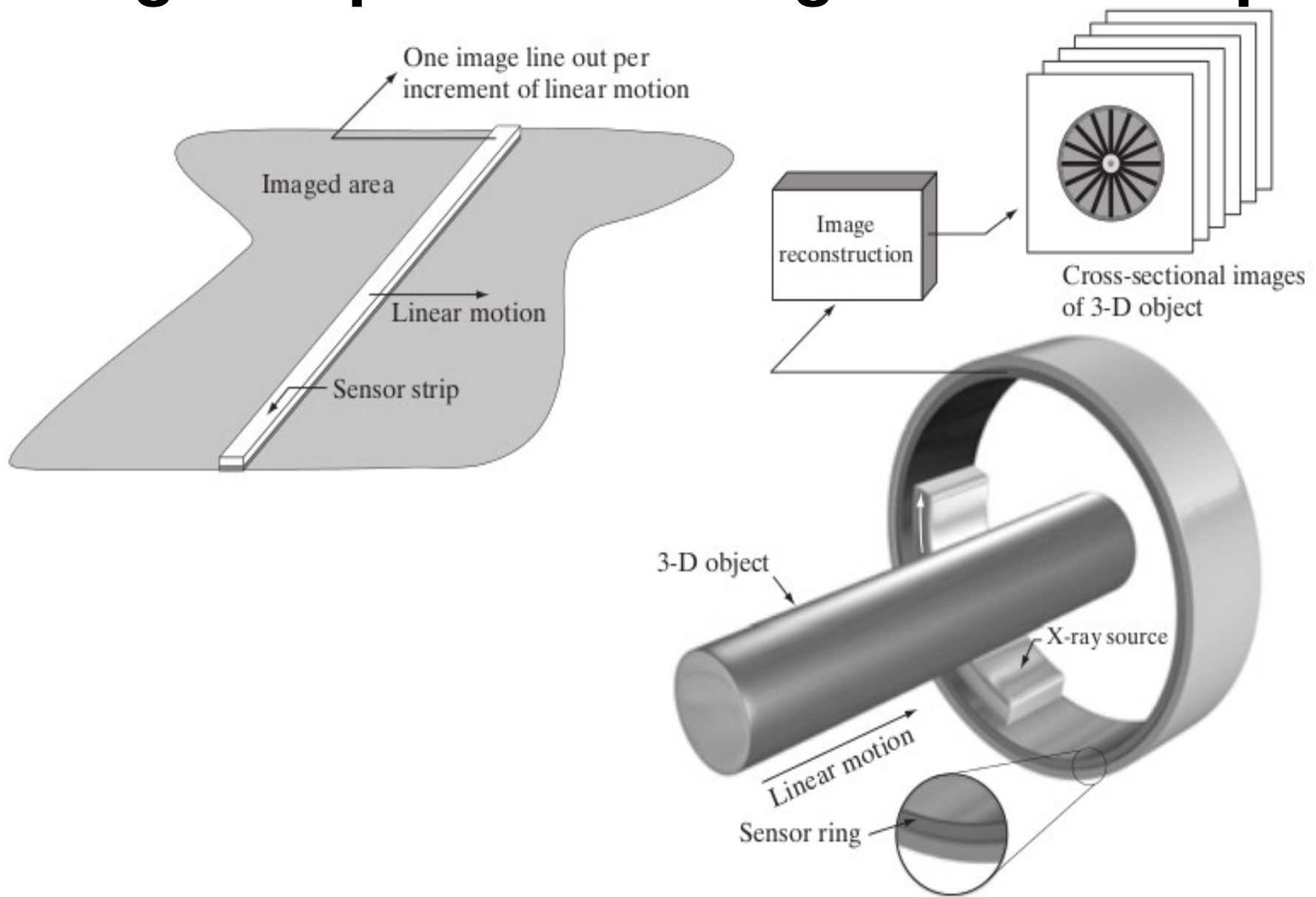


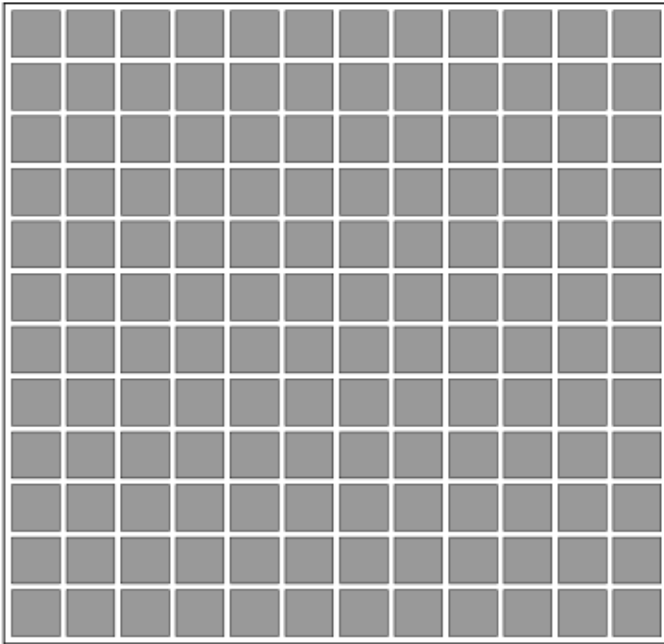
Image Acquisition Using Sensor Strips

- More frequently than single sensors consists of an in-line arrangement of sensors in the form of a *sensor strip*
- The strip provides imaging elements in one direction.
- Motion perpendicular to the strip provides imaging in the other direction
- Sensing devices with 4000 or more in-line sensors are possible.
- Used in most flat bed scanners, airborne applications

Image Acquisition Using Sensor Strips mounted in ring configuration

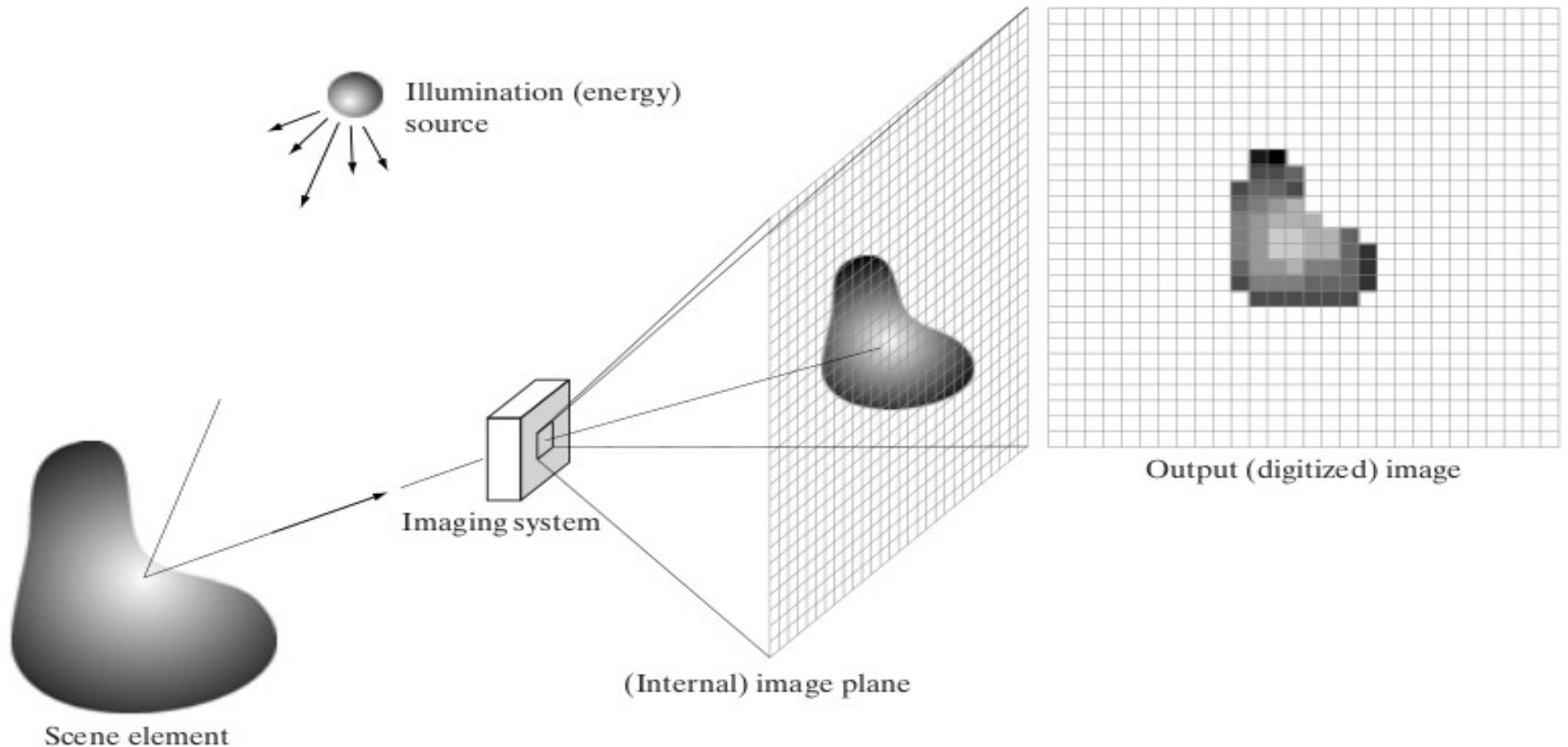
- Sensor strips mounted in a *ring configuration*
- A rotating X-ray/ similar source provides illumination
- Portion of the sensors opposite the source collect the X-ray energy that pass through the object
- Basis for medical and industrial computerized axial tomography (CAT)
- Output of the sensors must be processed by reconstruction algorithms whose objective is to transform the sensed data into meaningful cross-sectional images.
-
- Used in medical and industrial imaging to obtain cross-sectional (“slice”) images of 3-D objects
- Magnetic resonance imaging (MRI) and positron emission tomography (PET).

Image Acquisition Using Sensor Arrays

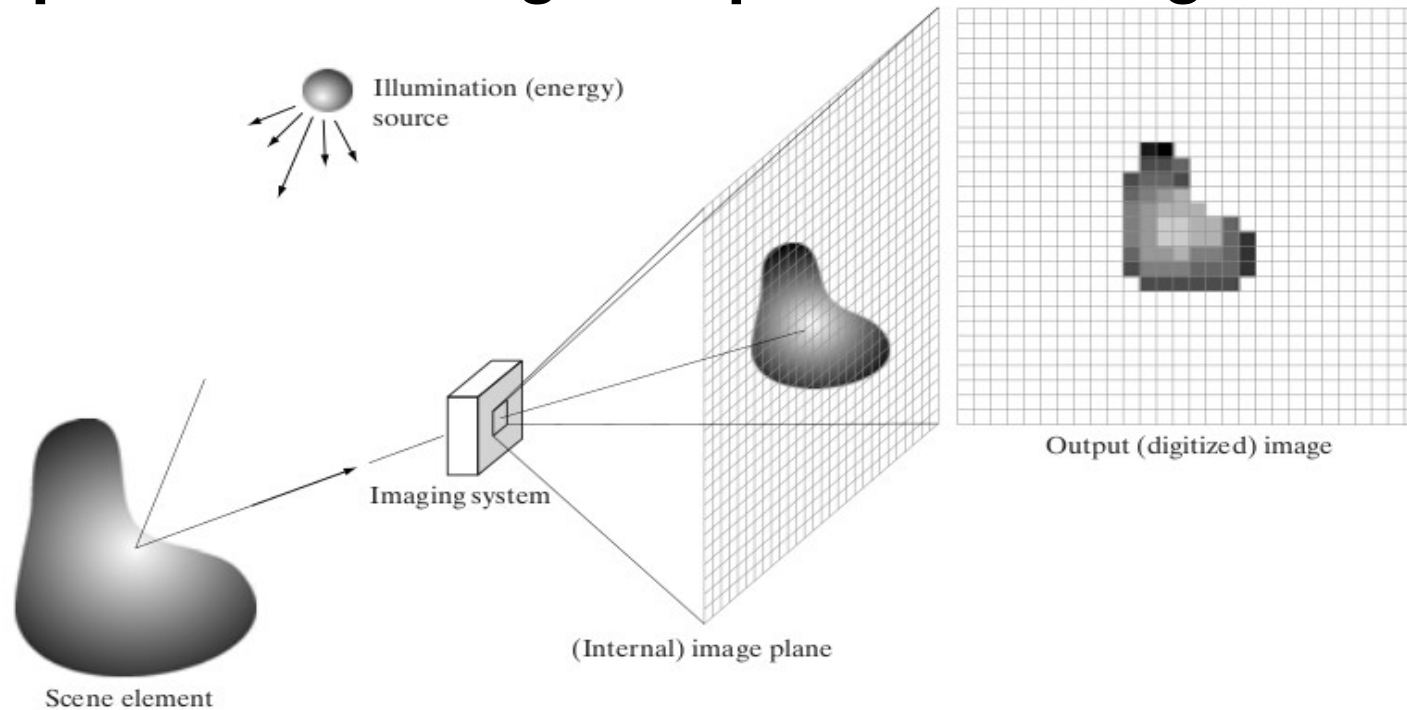


- Individual sensors arranged in the form of a 2-D array.
- Packaged in rugged arrays of $4000 * 4000$ elements or more
- CCD sensors are used widely in digital cameras and other light sensing instruments
- The response of each sensor is proportional to the integral of the light energy projected onto the surface of the sensor
- Noise reduction is achieved by letting the sensor integrate the input light signal over minutes or even hours
- Since array is 2D, complete image can be obtained and hence motion of elements is not necessary

Principle behind Image Acquisition Using Sensor Arrays



Principle behind Image Acquisition Using Sensor Arrays



- Collects the incoming energy and focus it onto an image plane
- Projects the viewed scene onto the lens focal plane, if the illumination is light
- The sensor array, which is coincident with the focal plane, produces outputs proportional to the integral of the light received at each sensor
- Digital and analog circuitry sweep these outputs and convert them to a video signal, which is then digitized by another section of the imaging system.
- Yielding the output, that is a digital image

A Simple Image Formation Model



A Simple Image Formation Model

```
octave:14> x=imread('gnu.jpg')
x =
255 255 251 255 252 251 219 203 206 228 253 255 254 255 253 255 255 250 255 255 249 255 255 249 238 243 248 255 252 255 254 252
250 251 255 222 167 182 167 178 137 139 155 208 249 244 255 255 253 255 252 253 255 237 175 168 163 175 188 179 230 254 245 255
255 253 195 174 183 53 73 128 157 140 187 209 133 224 250 253 251 254 255 255 156 191 191 139 128 105 63 139 184 191 255 251
255 191 186 119 16 170 165 184 208 199 181 190 204 255 255 254 255 254 255 249 231 185 192 188 201 180 174 75 60 186 213 255
250 165 195 0 185 167 255 247 254 255 252 238 201 186 203 238 254 255 234 198 207 236 242 255 255 255 221 191 54 132 168 248
195 240 69 69 178 252 248 255 253 253 209 182 218 173 175 178 203 176 180 182 190 205 195 251 254 255 251 191 137 44 218 209
183 237 55 97 176 255 253 255 255 212 186 113 10 0 6 52 169 110 108 30 3 51 183 202 255 246 255 254 135 22 222 189
175 245 37 114 203 253 255 251 219 173 80 3 0 3 0 3 79 251 181 158 22 2 33 171 214 255 252 255 146 21 231 170
182 236 56 118 180 255 255 230 170 95 1 4 14 32 16 82 134 252 226 178 69 4 1 61 173 240 254 242 143 31 227 177
189 239 64 68 168 226 219 161 115 51 144 180 234 206 221 255 227 183 176 228 247 167 192 70 90 178 218 177 134 38 242 167
194 223 113 4 142 181 155 72 113 243 255 240 135 42 233 250 166 214 186 195 87 187 235 245 60 61 162 164 7 79 216 207
251 176 184 16 4 34 22 80 255 254 247 176 213 200 222 198 219 199 247 211 121 220 220 252 201 7 6 6 16 168 171 244
253 197 203 175 43 8 0 208 249 255 250 245 249 215 118 77 183 221 246 228 85 59 212 255 255 48 18 48 177 198 216 255
255 255 183 189 229 146 102 252 255 251 253 216 236 229 151 79 225 253 255 244 130 90 227 250 255 182 107 232 180 194 246 255
255 255 253 216 153 157 164 255 255 255 255 170 247 250 251 225 221 250 249 255 241 153 176 255 255 249 143 183 237 255 255 249
253 253 255 250 255 217 196 255 252 255 222 111 255 254 249 255 254 255 254 253 255 195 134 166 206 255 216 155 252 255 247 255
255 255 249 255 255 150 235 253 253 232 147 231 237 255 255 251 223 148 179 183 245 255 127 193 187 180 176 190 255 255 255 251
254 249 255 254 226 157 255 219 163 87 248 255 247 251 255 253 140 150 201 230 182 227 247 220 164 250 255 255 255 250 254 255
255 255 250 255 121 178 120 112 238 136 244 255 255 231 214 255 232 218 104 156 253 238 255 254 174 242 255 255 248 255 252 253
255 255 250 251 190 210 216 149 251 155 231 248 255 255 243 207 230 248 251 236 246 252 255 223 155 254 250 247 255 251 255 254
250 255 255 252 255 253 229 152 255 236 160 255 255 254 254 214 186 255 255 253 255 250 189 121 248 255 255 255 255 252 255 255
255 252 251 255 248 255 255 135 251 254 117 209 247 249 255 255 166 244 247 255 255 255 251 191 211 255 250 250 255 252 255 250
254 255 255 255 255 249 252 129 255 251 253 111 223 255 250 251 239 165 164 171 166 129 95 136 253 247 255 255 250 255 255 255
255 249 255 254 252 255 255 162 227 255 250 145 221 246 255 254 255 249 243 246 248 255 203 226 255 249 255 250 255 255 255 253
255 255 253 253 255 255 249 241 163 252 255 235 121 129 253 255 255 242 182 125 98 131 163 247 255 255 255 255 255 255 255
255 255 255 254 255 255 254 250 194 206 250 255 237 105 153 202 248 255 209 207 130 252 245 255 255 255 255 255 255 255 255
254 255 255 255 254 255 255 255 253 194 232 255 233 93 35 63 253 253 255 237 197 213 255 254 255 255 255 255 255 255 255 255
254 255 255 255 253 253 254 255 255 251 211 245 255 232 83 66 172 254 213 117 142 249 250 250 255 255 255 255 255 255 255 255
255 255 255 255 255 254 253 253 255 254 255 252 250 234 111 193 18 150 17 10 220 252 255 255 255 255 255 255 255 255 255 255
255 254 254 255 255 255 255 254 255 252 252 254 255 238 213 169 38 40 4 0 38 140 241 250 255 255 255 255 255 255 255 255
255 253 254 255 255 255 255 255 251 255 254 254 255 251 255 223 191 38 33 62 27 187 243 255 255 255 255 255 255 255 255 255
255 254 254 255 255 253 254 255 255 253 255 255 255 255 250 255 252 237 234 229 244 255 255 250 255 255 255 255 255 255 255
```

Array reading the pixel values of the 32x32 gnu image

A Simple Image Formation Model

- Digital images by two-dimensional functions of the form
 $f(x, y)$
- The value or amplitude of f at spatial coordinates (x, y) is a positive scalar quantity
- When an image is generated from a physical process, its values are proportional to energy radiated by a physical source
- $f(x, y)$ must be nonzero and finite i.e.,

$$0 < f(x, y) < \infty$$

A Simple Image Formation Model

$f(x, y)$ may be characterized by two components:

- (1) The amount of source illumination incident on the scene being viewed – *Illumination component* - $i(x, y)^*$
- (2) the amount of illumination reflected by the objects in the scene – *Reflectance component* – $r(x, y)^{**}$

$$f(x, y) = i(x, y)r(x, y)$$

* $0 < i(x, y) < \infty$

** $0 < r(x, y) < \infty$

A Simple Image Formation Model

Gray Level (l)

$$l = f(x_0, y_0)$$

$$L_{min} \leq l \leq L_{max}$$

In theory, the only requirement on L_{min} is that it be positive, and on L_{max} that it be finite.

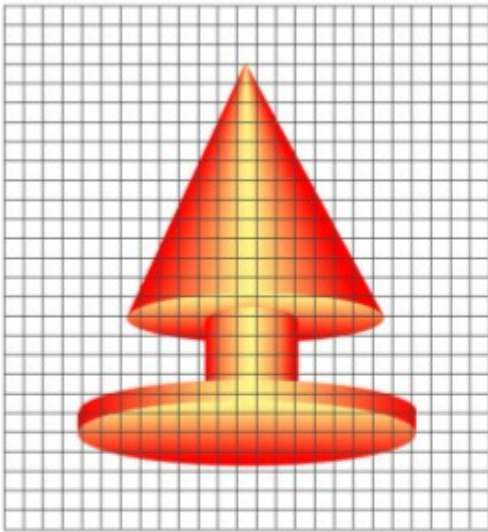
*In practice, $L_{min} = i_{min} * r_{min}$ and $L_{max} = i_{max} * r_{max}$*

A Simple Image Formation Model

Gray Level (l)

- The interval $[L_{\min}, L_{\max}]$ is called the gray scale.
- This interval takes values $[0, L-1]$, where $l=0$ is considered black and $l=(L-1)$ is considered white on the gray scale.
- All intermediate values are shades of gray varying from black to white.

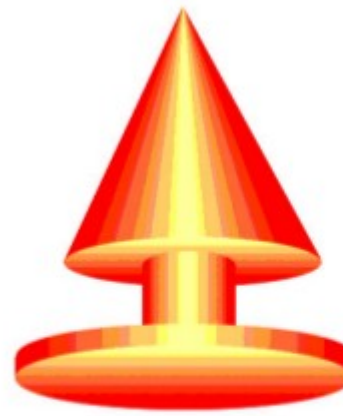
Image Sampling and Quantization



real image



sampled



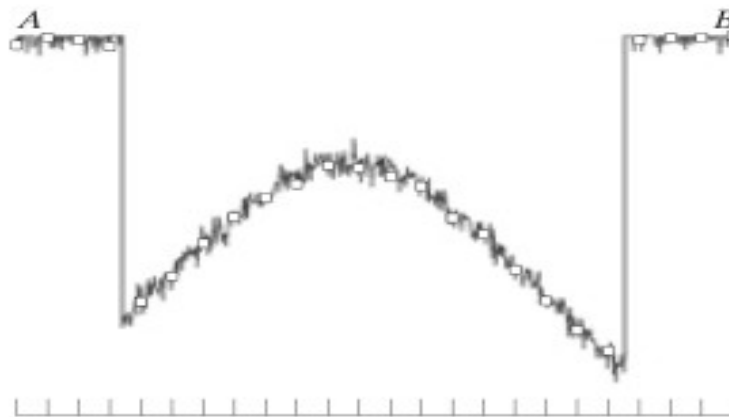
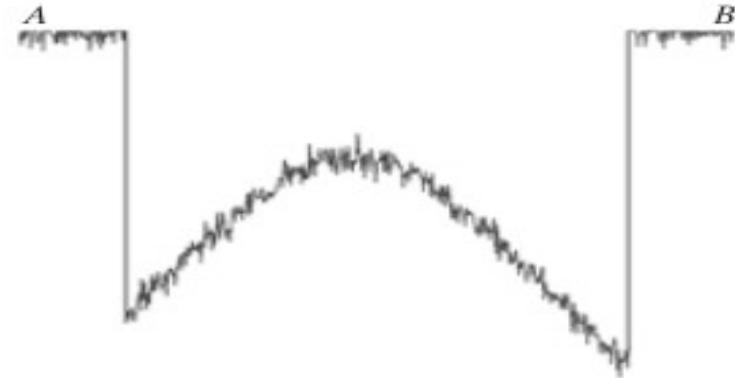
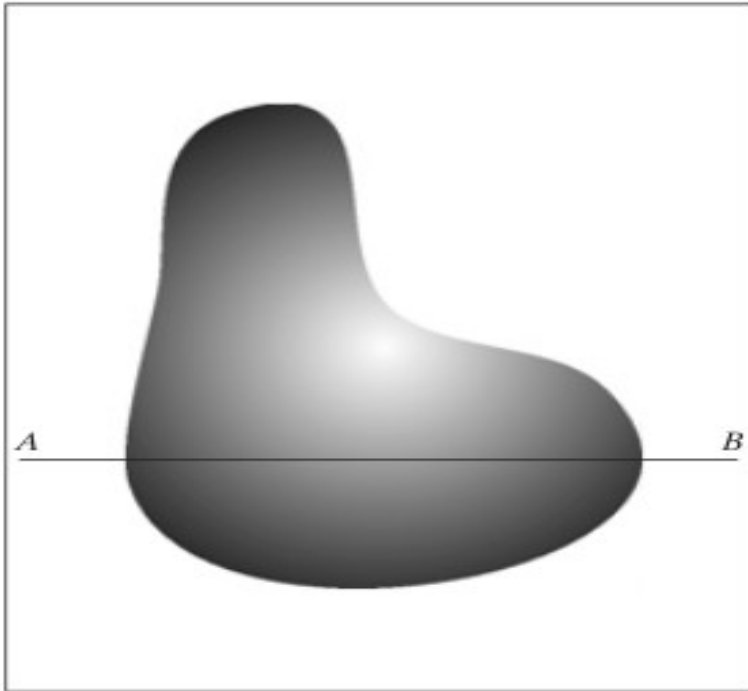
quantized



sampled &
quantized

Objective is to generate digital images from sensed data for, output of most sensors is a continuous voltage waveform

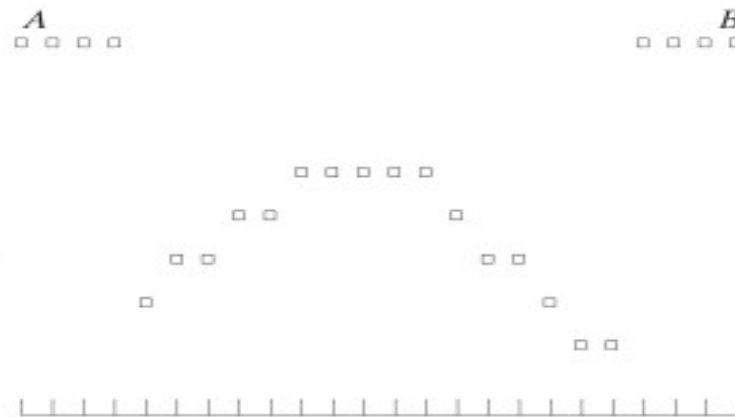
Basic Concepts in Sampling and Quantization



Sampling



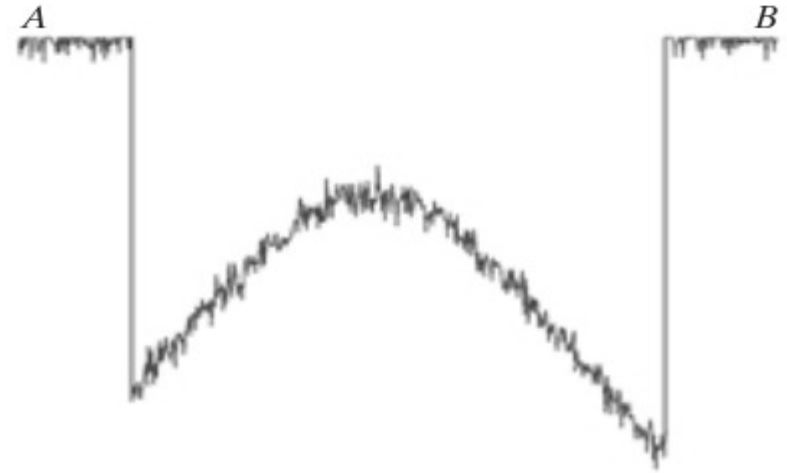
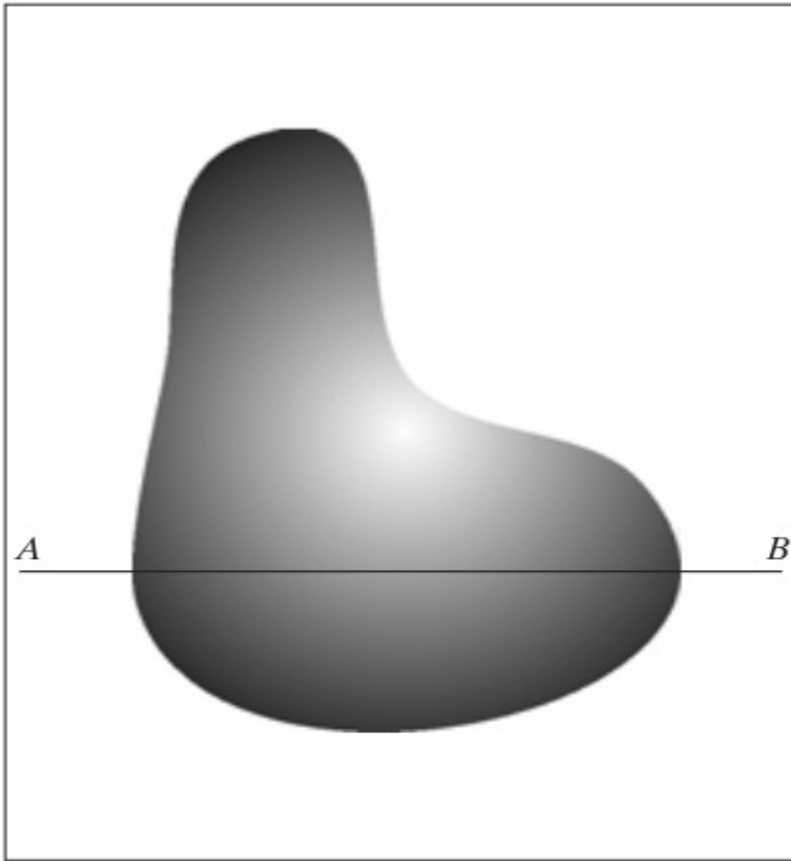
Quantization



Sampling and Quantization

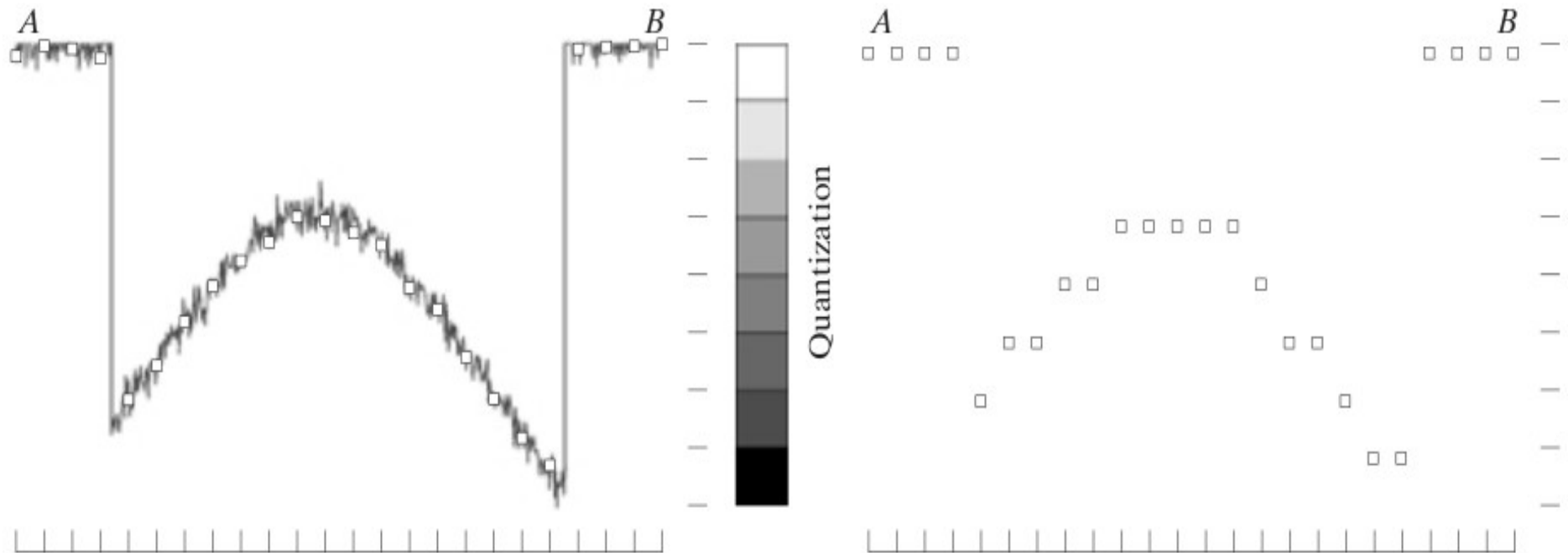
- An image may be continuous with respect to the x- and y-coordinates, and also in amplitude.
- To convert it to digital form, we have to sample the function in both coordinates and in amplitude.
- Digitizing the coordinate values is called ***sampling***.
- Digitizing the amplitude values is called ***quantization***.

Sampling and Quantization



The one-dimensional function shown in the figure above is a plot of amplitude (gray level) values of the continuous image along the line segment AB. The random variations are due to image noise.

Sampling and Quantization

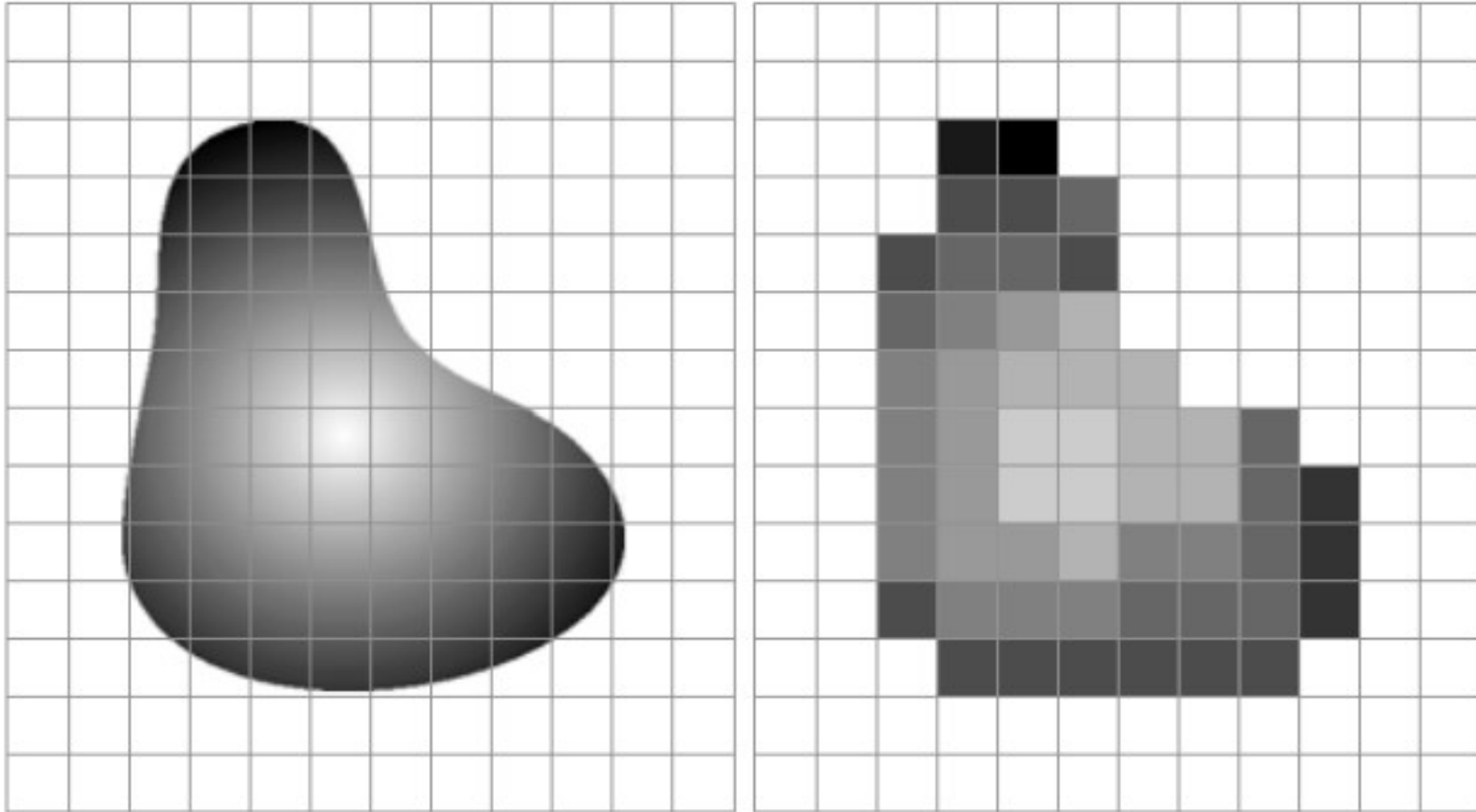


The values of the samples still span (vertically) a continuous range of gray-level values. In order to form a digital function, the gray-level values also must be converted (quantized) into discrete quantities

Sampling and Quantization in relation to the different methods of image acquisition

- *Single sensor* coupled with linear mechanical motion sensing : sampling is accomplished by selecting the number of individual mechanical increments at which we activate the sensor to collect data
- When a *sensing strip* is used for image acquisition, the number of sensors in the strip establishes the sampling limitations in one image direction. Mechanical motion in the other direction can be controlled more accurately

Sampling and Quantization in relation to the different methods of image acquisition



- When a sensing array is used for image acquisition, there is no motion and the number of sensors in the array establishes the limits of sampling in both directions.

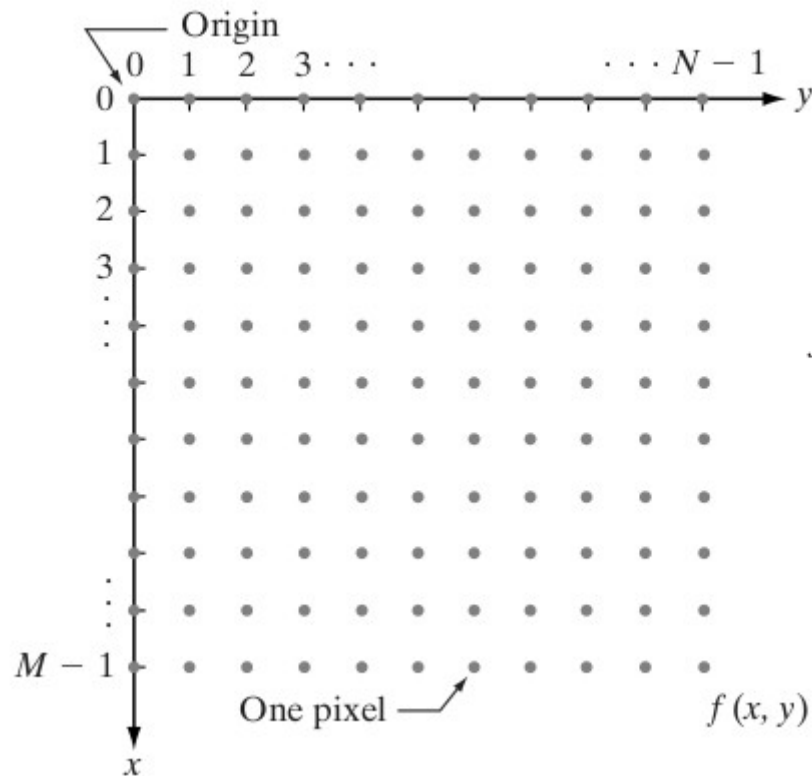
Representing Digital Images

```
octave:14> x=imread('gnu.jpg')
x =

255 255 251 255 252 251 219 203 206 228 253 255 254 255 253 255 255 250 255 255 249 255 255 249 238 243 248 255 252 255 254 252
250 251 255 222 167 182 167 178 137 139 155 208 249 244 255 255 253 255 252 253 255 237 175 168 163 175 188 179 230 254 245 255
255 253 195 174 183 53 73 128 157 140 187 209 133 224 250 253 251 254 255 255 156 191 191 139 128 105 63 139 184 191 255 251
255 191 186 119 16 170 165 184 208 199 181 190 204 255 255 254 255 254 255 249 231 185 192 188 201 180 174 75 60 186 213 255
250 165 195 0 185 167 255 247 254 255 252 238 201 186 203 238 254 255 234 198 207 236 242 255 255 255 221 191 54 132 168 248
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253 197 203 175 43 8 0 208 249 255 250 245 249 215 118 77 183 221 246 228 85 59 212 255 255 48 18 48 177 198 216 255
255 255 183 189 229 146 102 252 255 251 253 216 236 229 151 79 225 253 255 244 130 90 227 250 255 182 107 232 180 194 246 255
255 255 253 216 153 157 164 255 255 255 255 170 247 250 251 225 221 250 249 255 241 153 176 255 255 249 143 183 237 255 255 249
253 253 255 250 255 217 196 255 252 255 222 111 255 254 249 255 254 255 254 253 255 195 134 166 206 255 216 155 252 255 247 255
255 255 249 255 255 150 235 253 253 232 147 231 237 255 255 251 223 148 179 183 245 255 127 193 187 180 176 190 255 255 255 251
254 249 255 254 226 157 255 219 163 87 248 255 247 251 255 253 140 150 201 230 182 227 247 220 164 250 255 255 255 250 254 255
255 255 250 255 121 178 120 112 238 136 244 255 255 231 214 255 232 218 104 156 253 238 255 254 174 242 255 255 248 255 252 253
255 255 250 251 190 210 216 149 251 155 231 248 255 255 243 207 230 248 251 236 246 252 255 223 155 254 250 247 255 251 255 254
250 255 255 252 255 253 229 152 255 236 160 255 255 254 254 214 186 255 255 253 255 250 189 121 248 255 255 255 255 252 255 255
255 252 251 255 248 255 255 135 251 254 117 209 247 249 255 255 166 244 247 255 255 255 251 191 211 255 250 250 255 252 255 250
254 255 255 255 255 249 252 129 255 251 253 111 223 255 250 251 239 165 164 171 166 129 95 136 253 247 255 255 250 255 255 255
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255 255 255 254 255 255 254 250 194 206 250 255 237 105 153 202 248 255 209 207 130 252 245 255 255 255 255 255 255 255 255
254 255 255 255 254 255 255 255 253 194 232 255 233 93 35 63 253 253 255 237 197 213 255 254 255 255 255 255 255 255 255
254 255 255 255 253 253 254 255 255 251 211 245 255 232 83 66 172 254 213 117 142 249 250 250 255 255 255 255 255 255 255
255 255 255 255 255 254 253 253 255 254 255 252 250 234 111 193 18 150 17 10 220 252 255 255 255 255 255 255 255 255 255
255 254 254 255 255 255 255 254 255 252 252 254 255 238 213 169 38 40 4 0 38 140 241 250 255 255 255 255 255 255 255
255 253 254 255 255 255 255 255 251 255 254 254 255 251 255 223 191 38 33 62 27 187 243 255 255 255 255 255 255 255 255
255 254 254 255 255 253 254 255 255 253 255 255 255 250 255 252 237 234 229 244 255 255 250 255 255 255 255 255 255 255
```

- The result of sampling and quantization is a matrix of real numbers.

Representing Digital Images: Different notations



$$f(x, y) = \begin{bmatrix} f(0, 0) & f(0, 1) & \cdots & f(0, N - 1) \\ f(1, 0) & f(1, 1) & \cdots & f(1, N - 1) \\ \vdots & \vdots & & \vdots \\ f(M - 1, 0) & f(M - 1, 1) & \cdots & f(M - 1, N - 1) \end{bmatrix}.$$

$$\mathbf{A} = \begin{bmatrix} a_{0,0} & a_{0,1} & \cdots & a_{0,N-1} \\ a_{1,0} & a_{1,1} & \cdots & a_{1,N-1} \\ \vdots & \vdots & & \vdots \\ a_{M-1,0} & a_{M-1,1} & \cdots & a_{M-1,N-1} \end{bmatrix}.$$

Also can be denoted in more formal mathematical notations using **Z** and **R**

Representing Digital Images: Different notations

The number, b , of bits required to store a digitized image is

$$b = M \times N \times k.$$

When $M = N$, this equation becomes

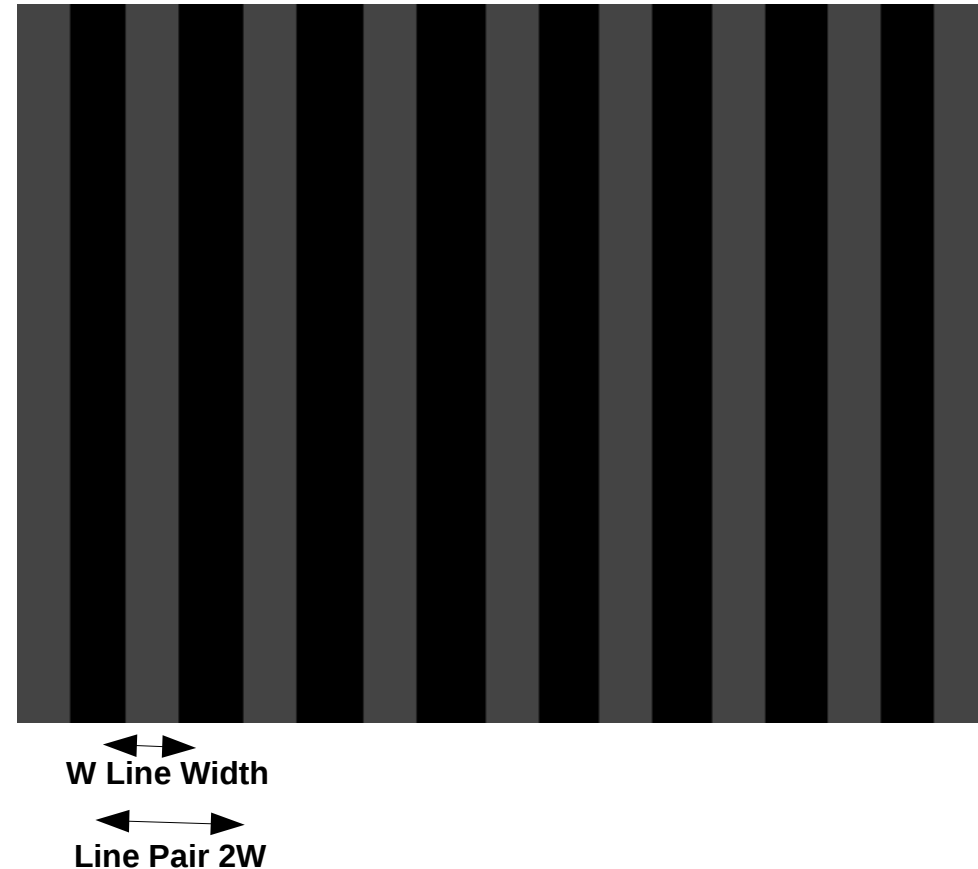
$$b = N^2 k.$$

Number of storage bits for various values of N and k .

N/k	1 ($L = 2$)	2 ($L = 4$)	3 ($L = 8$)	4 ($L = 16$)	5 ($L = 32$)	6 ($L = 64$)	7 ($L = 128$)	8 ($L = 256$)
32	1,024	2,048	3,072	4,096	5,120	6,144	7,168	8,192
64	4,096	8,192	12,288	16,384	20,480	24,576	28,672	32,768
128	16,384	32,768	49,152	65,536	81,920	98,304	114,688	131,072
256	65,536	131,072	196,608	262,144	327,680	393,216	458,752	524,288
512	262,144	524,288	786,432	1,048,576	1,310,720	1,572,864	1,835,008	2,097,152
1024	1,048,576	2,097,152	3,145,728	4,194,304	5,242,880	6,291,456	7,340,032	8,388,608
2048	4,194,304	8,388,608	12,582,912	16,777,216	20,971,520	25,165,824	29,369,128	33,554,432
4096	16,777,216	33,554,432	50,331,648	67,108,864	83,886,080	100,663,296	117,440,512	134,217,728
8192	67,108,864	134,217,728	201,326,592	268,435,456	335,544,320	402,653,184	469,762,048	536,870,912

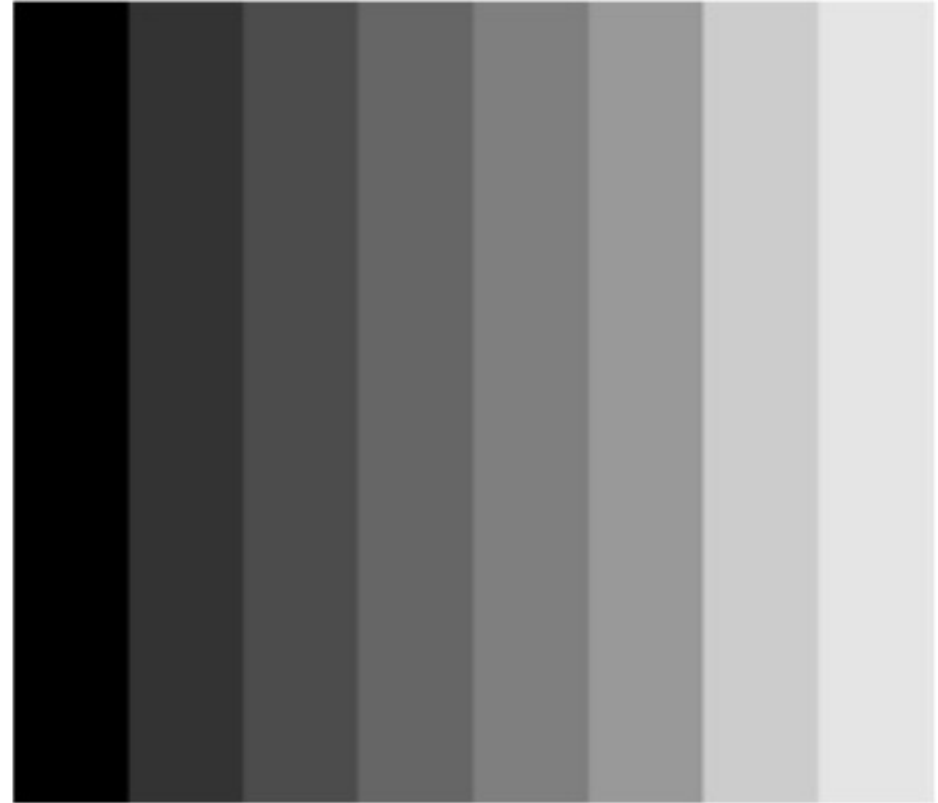
Spatial and Gray-Level Resolution

- ***Spatial resolution*** is the smallest discernible detail in an image
- *Sampling* is the principal factor determining the spatial resolution of an image
- Resolution is the smallest number of discernible ***line pairs*** per unit distance
- Ex: 100 line pairs per millimeter



Spatial and Gray-Level Resolution

- ***Gray-level resolution*** refers to the smallest discernible change in gray level (which btw is highly subjective)
- Considerable discretion can be accomplished regarding the number of samples used to generate a digital image, but this is not true for the number of gray levels
- The most common number is 8 bits, with 16 bits being used in some applications where enhancement of specific gray-level ranges is necessary.



Subsampling of a Gray-scale image

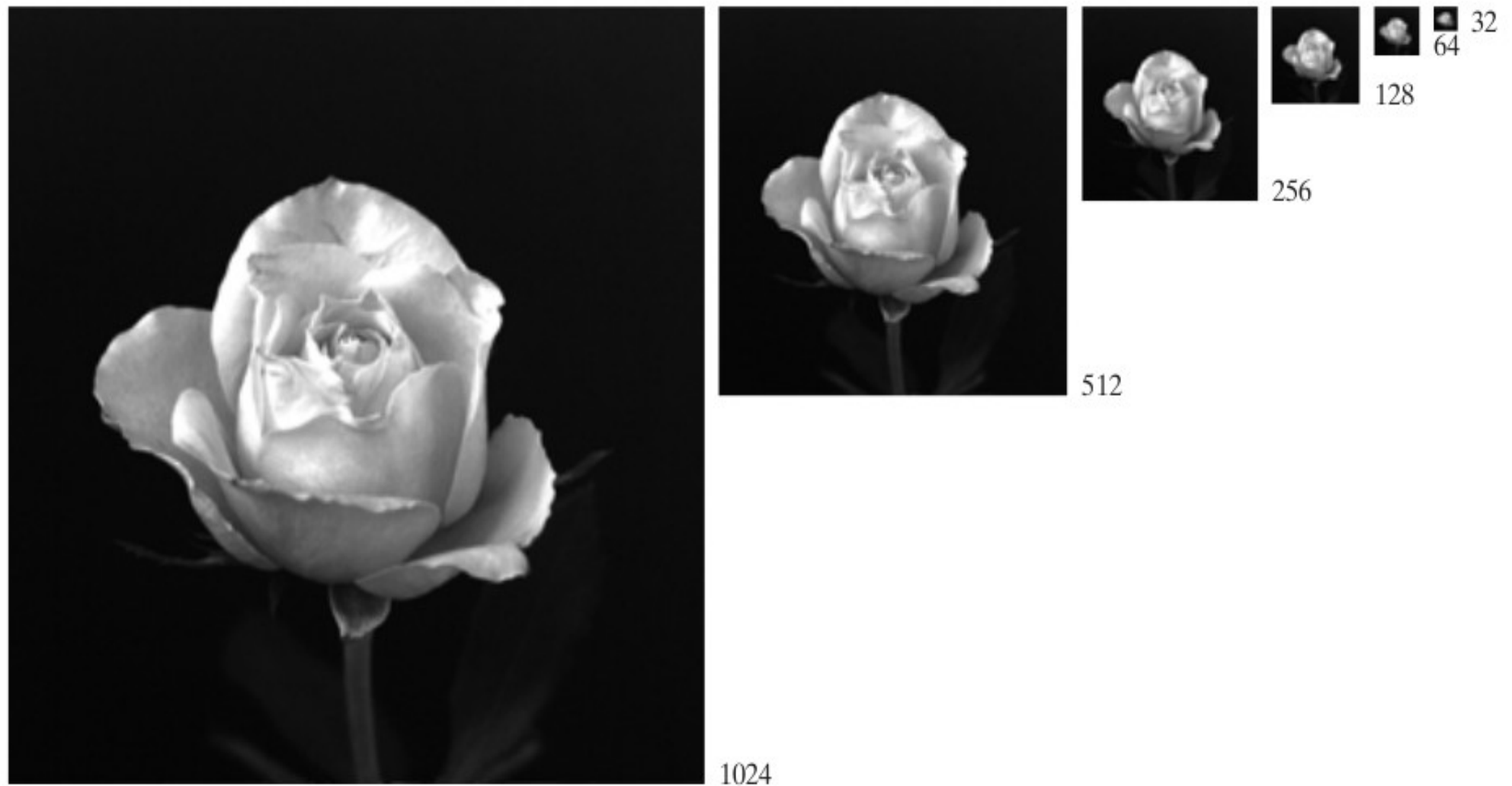
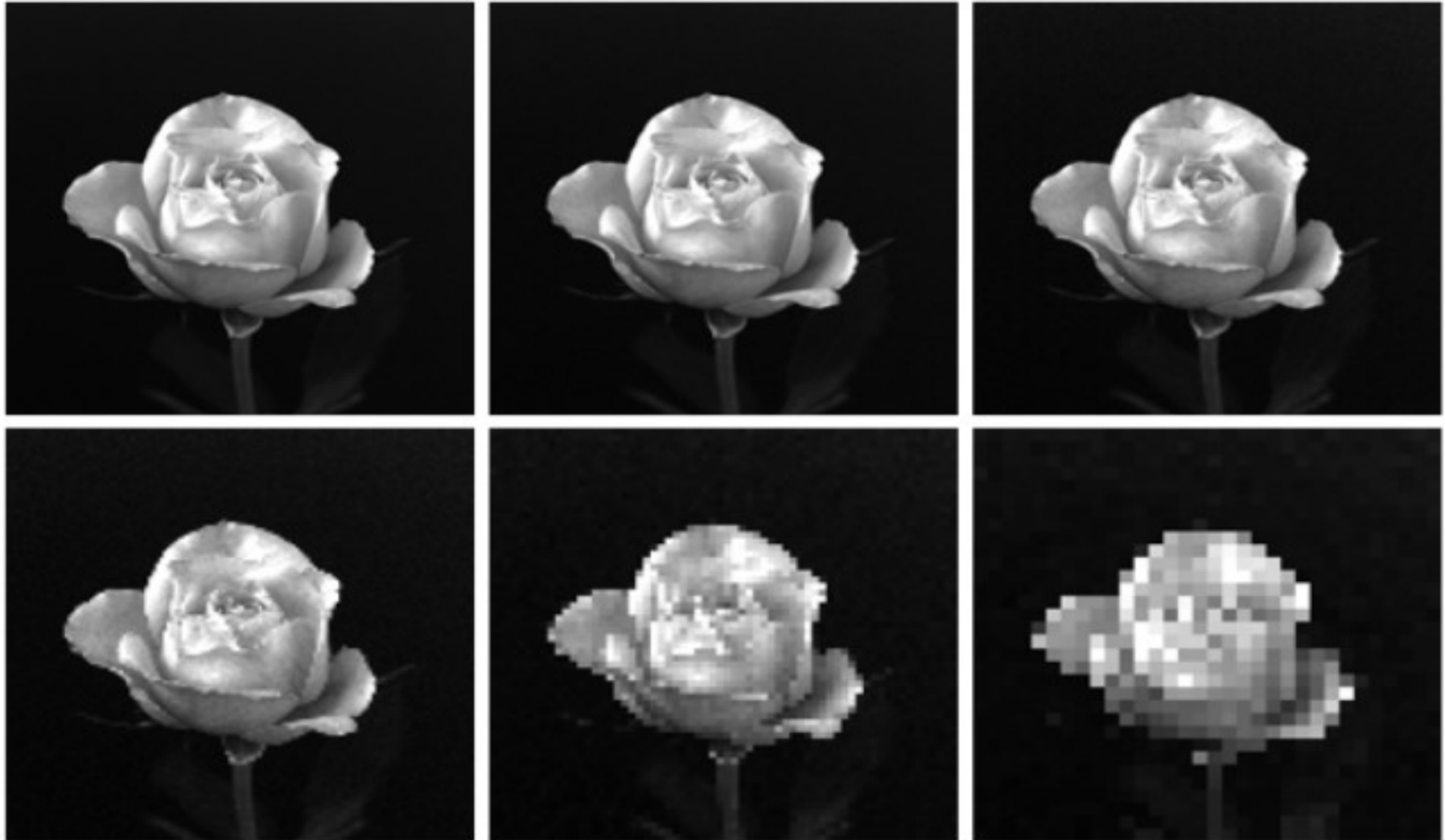


FIGURE 2.19 A 1024×1024 , 8-bit image subsampled down to size 32×32 pixels. The number of allowable gray levels was kept at 256.

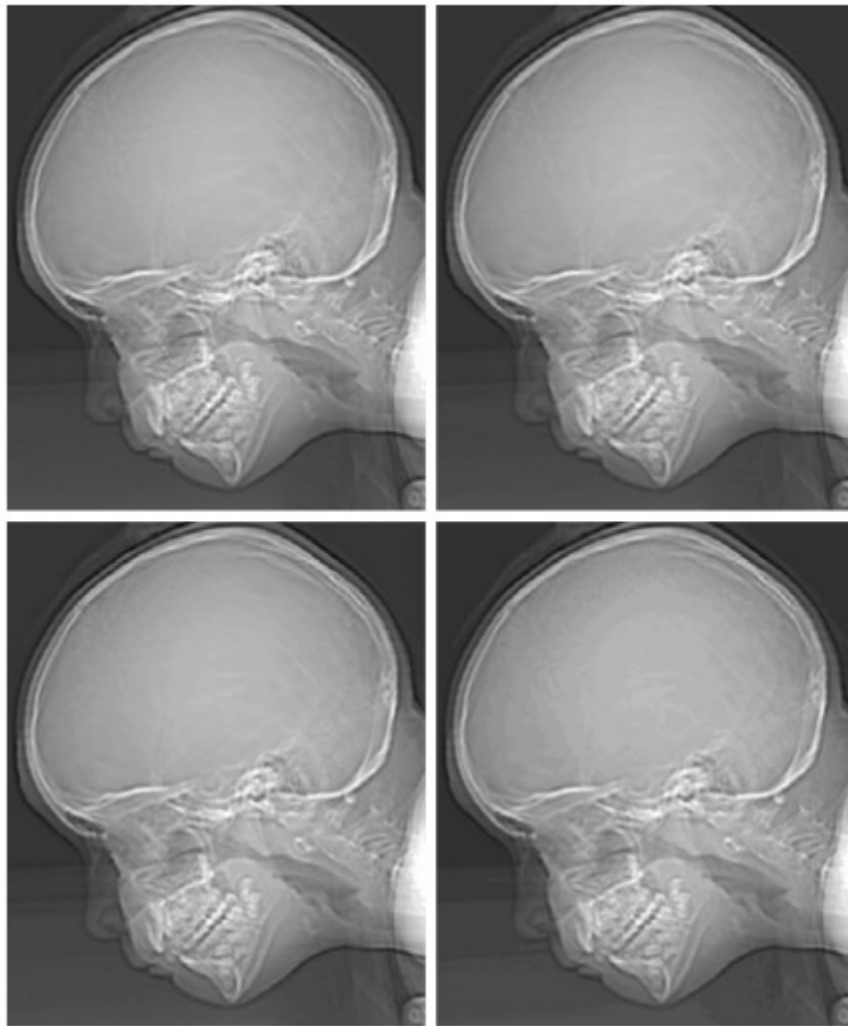
Resampling of a Gray-scale image



a	b	c
d	e	f

FIGURE 2.20 (a) 1024×1024 , 8-bit image. (b) 512×512 image resampled into 1024×1024 pixels by row and column duplication. (c) through (f) 256×256 , 128×128 , 64×64 , and 32×32 images resampled into 1024×1024 pixels.

Typical effects of varying the number of gray levels in a digital image.



a b
c d

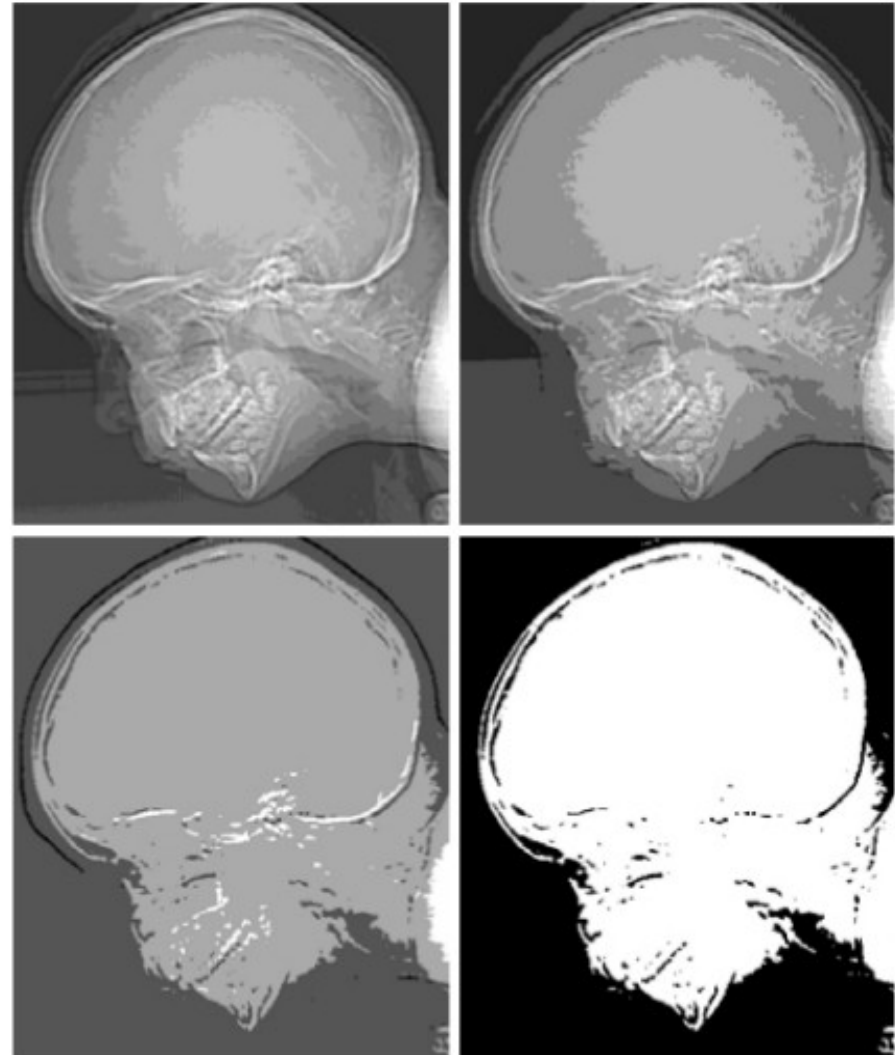
FIGURE 2.21

(a) 452×374 , 256-level image. (b)–(d) Image displayed in 128, 64, and 32 gray levels, while keeping the spatial resolution constant.

e f
g h

FIGURE 2.21

(Continued) (e)–(g) Image displayed in 16, 8, 4, and 2 gray levels. (Original courtesy of Dr. David R. Pickens, Department of Radiology & Radiological Sciences, Vanderbilt University Medical Center.)



Crude rule of thumb for sampling and quantization

Assuming powers of 2 for convenience, images of size 256×256 pixels and 64 gray levels are about the smallest images that can be expected to be reasonably free of objectionable sampling checkerboards and false contouring

Effects produced on image quality by varying N and k independently and simultaneously



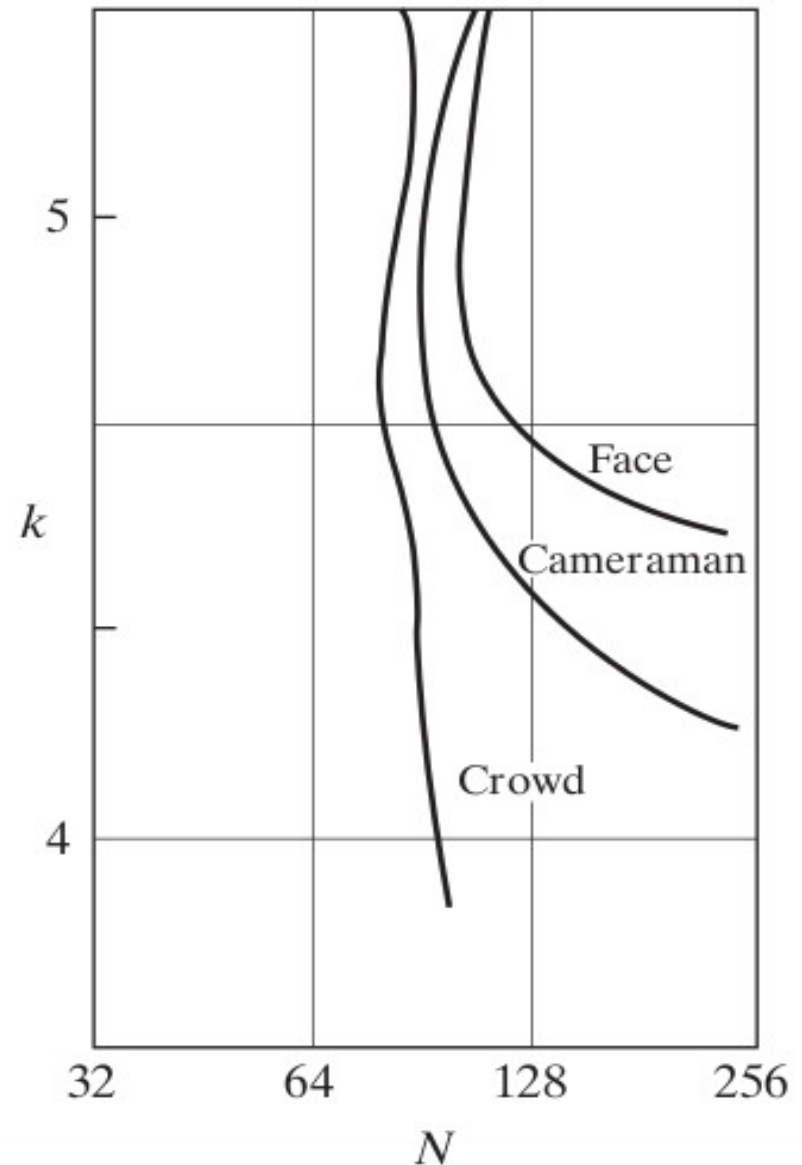
(a) Image with a low level of detail. (b) Image with a medium level of detail. (c) Image with a relatively large amount of detail

Isopreference curves

- An early study by Huang [1965] attempted to quantify experimentally the effects on image quality produced by varying N and k simultaneously
- Three sets of images : With little, intermediate and large amount of detail
- Sets of these three types of images were generated by varying N and k
- Observers were then asked to rank them according to their subjective quality.

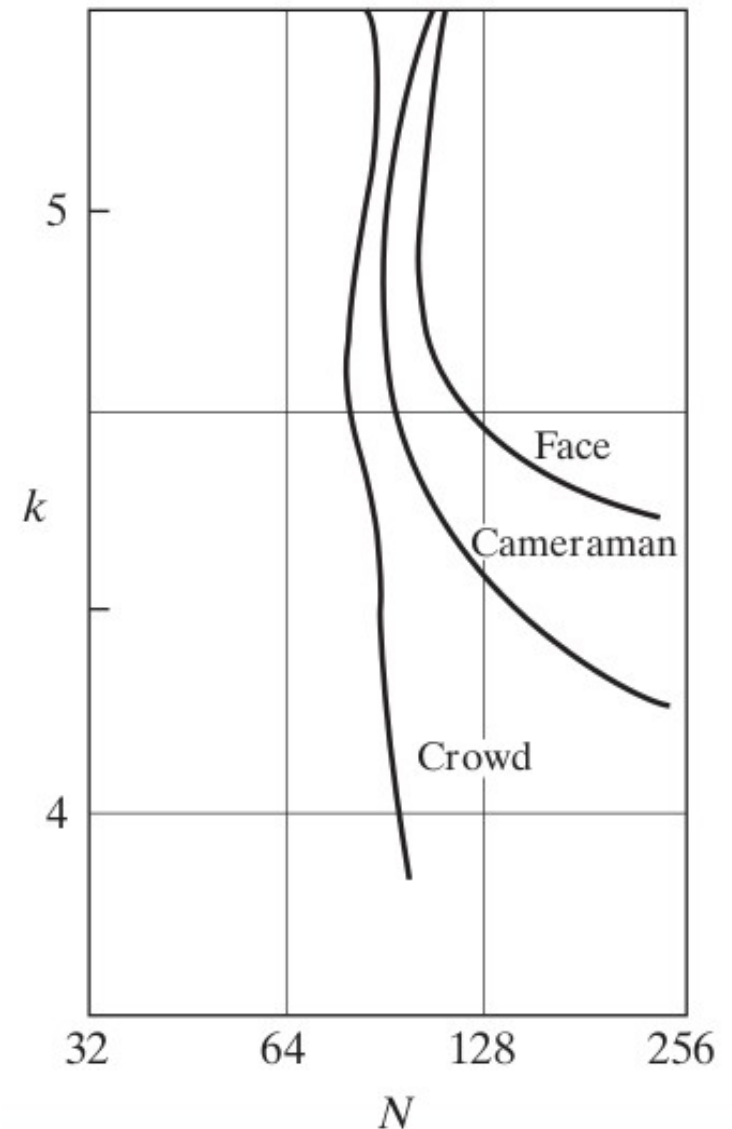
Isopreference curves

- Each point in the Nk -plane represents an image having values of N and k equal to the coordinates of that point
- Points lying on an isopreference curve correspond to images of equal subjective quality
- It was found that the *isopreference curves* tended to shift *right* and *upward*, but their shapes in each of the three image categories were *similar*



Inferences from the Isopreference curves

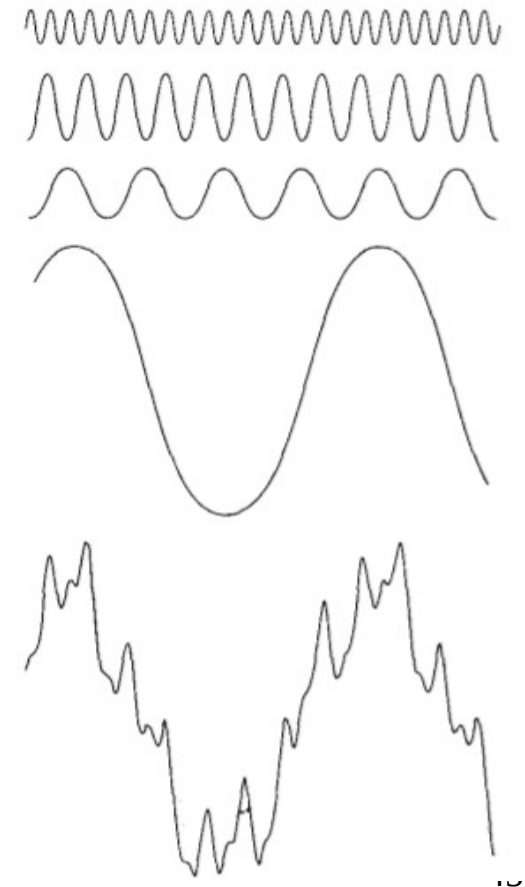
- A shift up and right in the curves simply means larger values for N and k , which implies **better picture quality**.
- Curves tend to become more *vertical* as the *detail in the image increases*.
- For images with a large amount of detail only a few gray levels may be needed
- This indicates that, for a fixed value of N , the perceived quality for this type of image is nearly independent of the number of gray levels used
- Perceived quality in the other two image categories remained the same in some intervals in which the spatial resolution was increased, but the number of gray levels actually decreased.
- A decrease in k tends to increase the apparent contrast of an image, a visual effect that humans often perceive as improved quality in an image.



Fourier Series and Fourier Transforms

$$f(x) = \frac{a_0}{2} + a_1 \cos(x) + a_2 \cos(2x) + a_3 \cos(3x) + \dots + a_n \cos(nx) \\ + b_1 \sin(x) + b_2 \sin(2x) + b_3 \sin(3x) + \dots + b_n \sin(nx)$$

$$F(u) = \int_{-\infty}^{\infty} f(x) e^{-j2\pi ux} dx$$



Fourier Series and Fourier Transforms

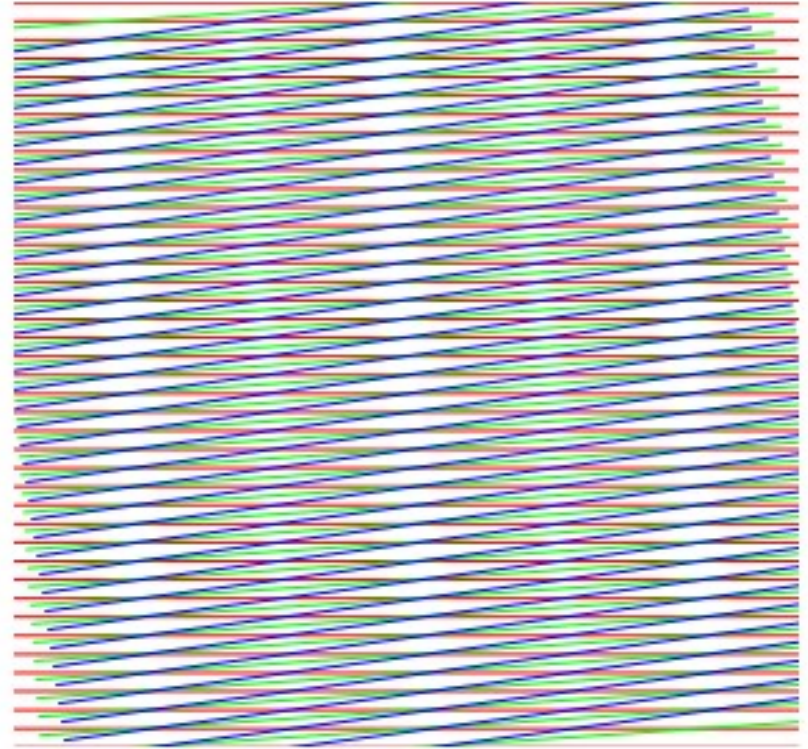
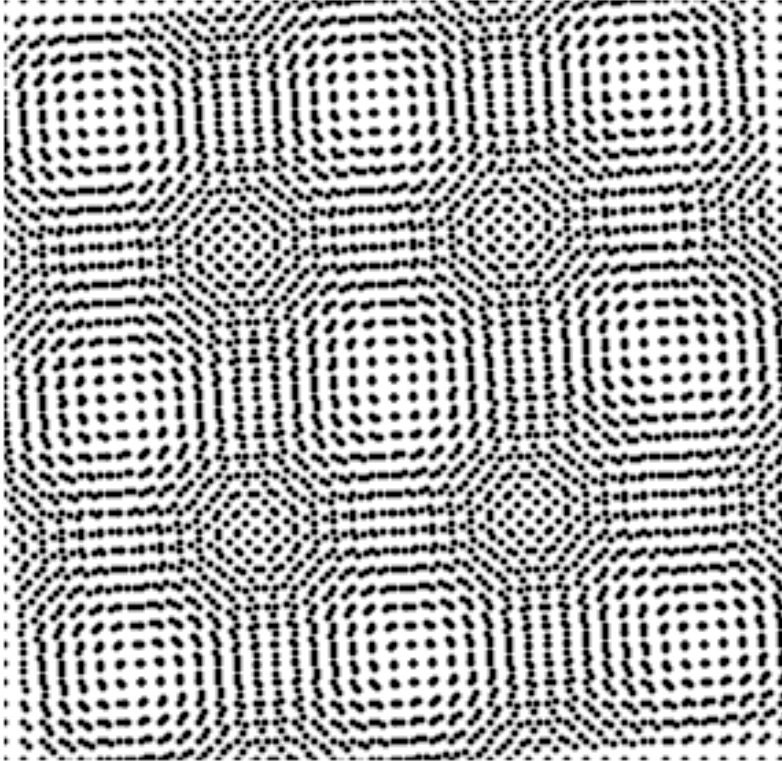
For any periodic function $f(x)$

$$f(x) = \frac{a_0}{2} + a_1 \cos(x) + a_2 \cos(2x) + a_3 \cos(3x) + \dots + a_n \cos(nx) \\ + b_1 \sin(x) + b_2 \sin(2x) + b_3 \sin(3x) + \dots + b_n \sin(nx)$$

For any non-periodic function $f(x)$

$$F(u) = \int_{-\infty}^{\infty} f(x) e^{-j2\pi ux} dx$$

Aliasing and Moiré* Patterns



*say mwahr

Aliasing and Moiré* Patterns

- Functions whose area under the curve is finite can be represented in terms of sines and cosines of various frequencies.
- The sine/cosine component with the highest frequency determines the highest “frequency content” of the function.
- If the function is band-limited, then Shannon's Sampling theorem can be applied
- If the function is *undersampled*, then a phenomenon called *aliasing* corrupts the sampled image
- The corruption is in the form of *additional frequency* components being introduced into the sampled function called *Aliased frequencies*

Note: *The sampling rate in images is the number of samples taken (in both spatial directions) per unit distance.*

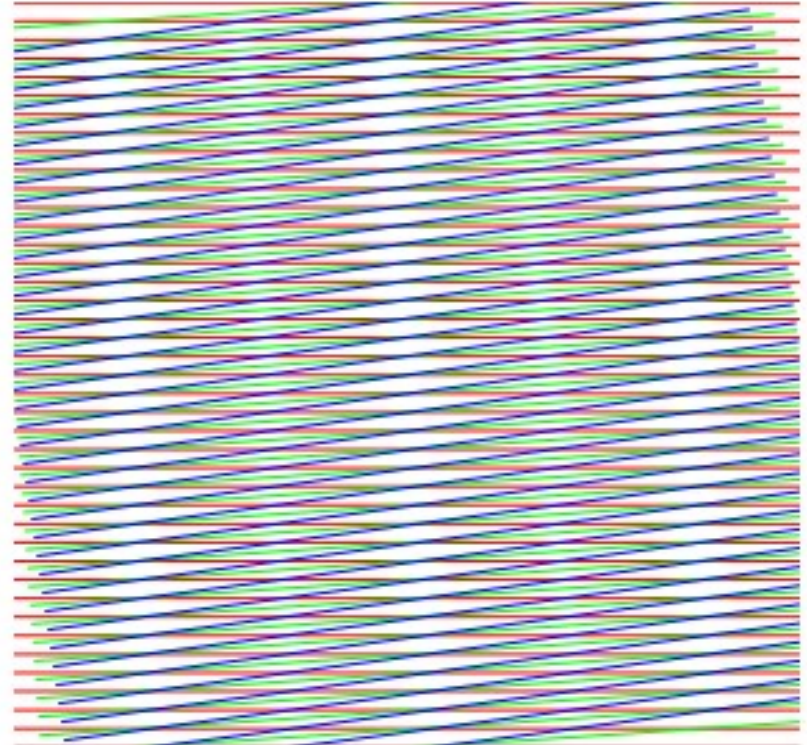
•

Practical constraints of sampling/ Paradox of sampling

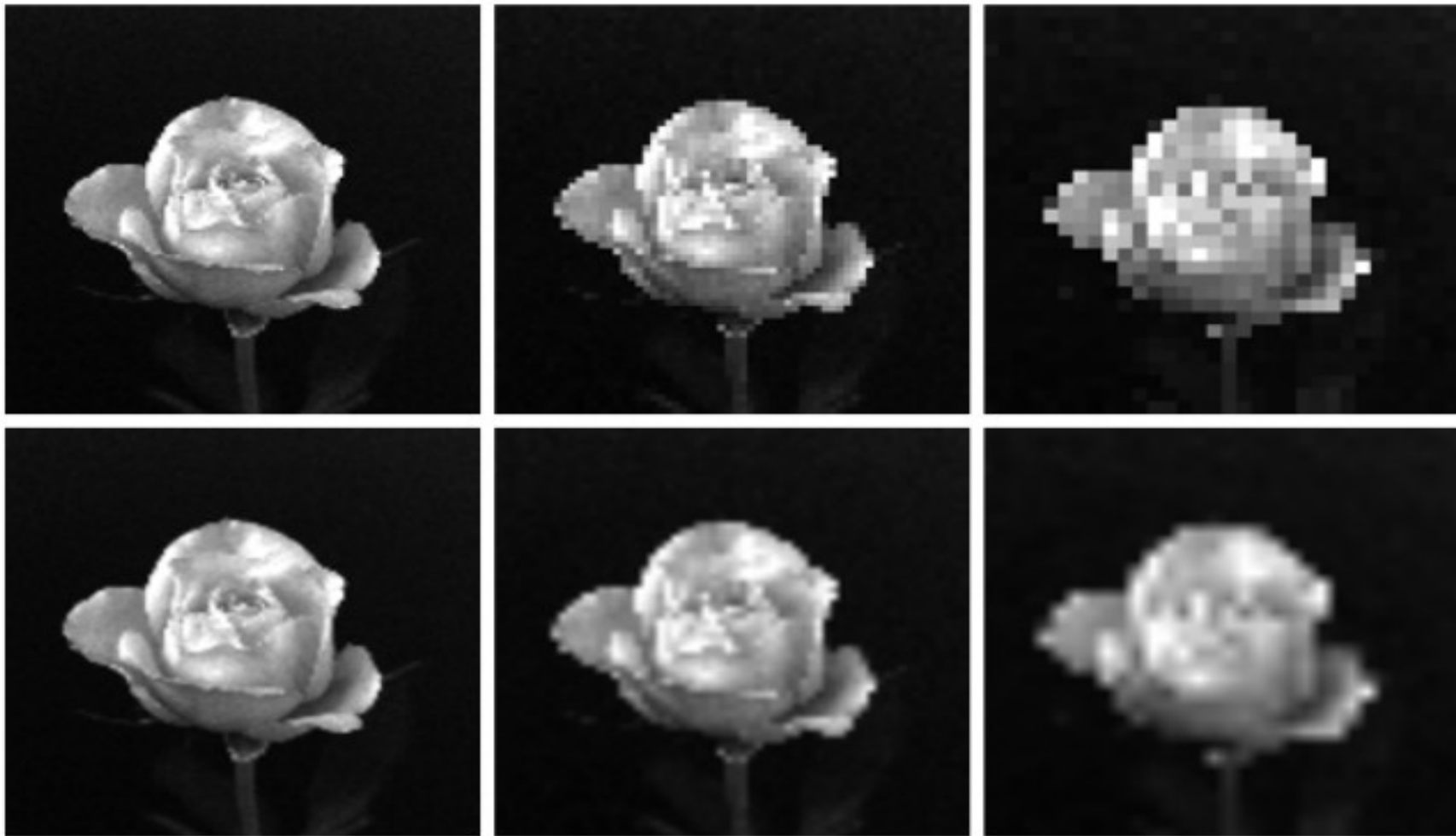
- In functions of infinite duration (hence finite highest frequencies) taking infinite samples is not feasible (but for one exception)
- We can work only with *finite samples of data*
- A function of unlimited duration is processed into a function of finite duration simply by multiplying the unlimited function by a “gating function” (|||lr to the Step functions)
- The gating function itself has frequency components that extend to infinity
- *The very act of limiting the duration of a band-limited function causes it to cease being band limited*
- Reducing the aliasing effects on an image is to reduce its high-frequency components by blurring the image *prior to sampling*
- The effect of *aliased frequencies* under the right conditions form the **Moiré Patterns**

What are Moiré* Patterns?

- **Special case:** a function of *infinite duration* can be sampled over a *finite interval* without violating the sampling theorem.
- With $f_s \geq 2f_h$, it is possible to recover the function from its samples provided that the sampling captures *exactly an integer number of periods* of the function.
- The Moiré pattern in the figure is caused by a breakup of the periodicity between the three overlapping images



Zooming and Shrinking Digital images



a	b	c
d	e	f

FIGURE 2.25 Top row: images zoomed from 128×128 , 64×64 , and 32×32 pixels to 1024×1024 pixels, using nearest neighbor gray-level interpolation. Bottom row: same sequence, but using bilinear interpolation.

Zooming and Shrinking Digital images

- Zooming -> Oversampling
- Shrinking -> Undersampling
- Key difference between Zooming and Shrinking, Sampling and Quantization
- Continuous and Digital images

Steps involved in Zooming

- Creation of new pixel locations
- Assignment of gray levels to those new locations.
- In order to perform gray-level assignment for any point in the overlay (newly inserted pixels), we look for the closest pixel in the original image and assign its gray level to the new pixel in the grid

Methods for zooming



140x140

Original

`a=imread('gnu.jpg')`

- Nearest Neighbour Interpolation
- Pixel Replication (special case of Nearest neighbour intr.)
- Bilinear Interpolation



210x210

Nearest neighbour Interpolation
`imresize(a,1.5,'nearest')`



210x210

Pixel replication
`imresize(a,1.5)`



210x210

Bilinear interpolation
`imresize(a,1.5,'bilinear')`

Methods for zooming:

Nearest Neighbour Interpolation

- Lay an overlay (grid) of the desired increased size of the image
- The spacing in the grid would be less than one pixel because we are fitting it over a smaller image
- Gray-level assignments to be made for any point in the overlay (newly inserted pixels)
- Look for the closest pixel in the original image and assign its gray level to the new pixel in the grid.
- When we are done with all points in the overlay grid, we simply expand it to the original specified size to obtain the zoomed image.
- Undesirable feature: *Checkerboard effect*, particularly objectionable at high magnification

Methods for zooming: Pixel Replication

- Special case of nearest neighbor interpolation
- Pixel replication is applicable when we want to increase the size of an image an integer number of times
- Duplication of rows and columns is done the required number of times to achieve the desired size
- The gray-level assignment of each pixel is predetermined by the fact that new locations are exact duplicates of old location

Methods for zooming: Bilinear Interpolation

- Overcomes the checkerboard effect from previous methods
- Uses four nearest neighbours of a point

Let (x', y') denote the coordinates of a point in the zoomed image (think of it as a point on the grid described previously), and let $v(x', y')$ denote the gray level assigned to it.

For bilinear interpolation, the assigned gray level is given by

$$v(x', y') = ax' + by' + cx'y' + d$$

where the four coefficients are determined from the four equations in four unknowns that can be written using the four nearest neighbors of point

Shrinking of digital images

- Image shrinking is performed in a similar manner as zooming
- The equivalent process of pixel replication is row-column deletion. For example, to shrink an image by one-half, we delete every other row and column.
- We can use the zooming grid analogy to visualize the concept of shrinking by a noninteger factor, except that we now expand the grid to fit over the original image, do gray-level nearest neighbor or bilinear interpolation, and then shrink the grid back to its original specified size.
- To reduce possible aliasing effects, it is a good idea to blur an image slightly before shrinking it

More on zooming and shrinking

- It is possible to use more neighbors for interpolation.
- Using more neighbors implies fitting the points with a more complex surface, which generally gives smoother results. This is an exceptionally important consideration in image generation for 3-D graphics and in medical image processing
- But the extra computational burden seldom is justifiable for general-purpose digital image zooming and shrinking, where bilinear interpolation generally is the method of choice.



210x210

Bilinear interpolation
`imresize(a,1.5,'bilinear')`



140x140

Original

`a=imread('gnu.jpg')`

<http://ece756.wikispaces.com/>



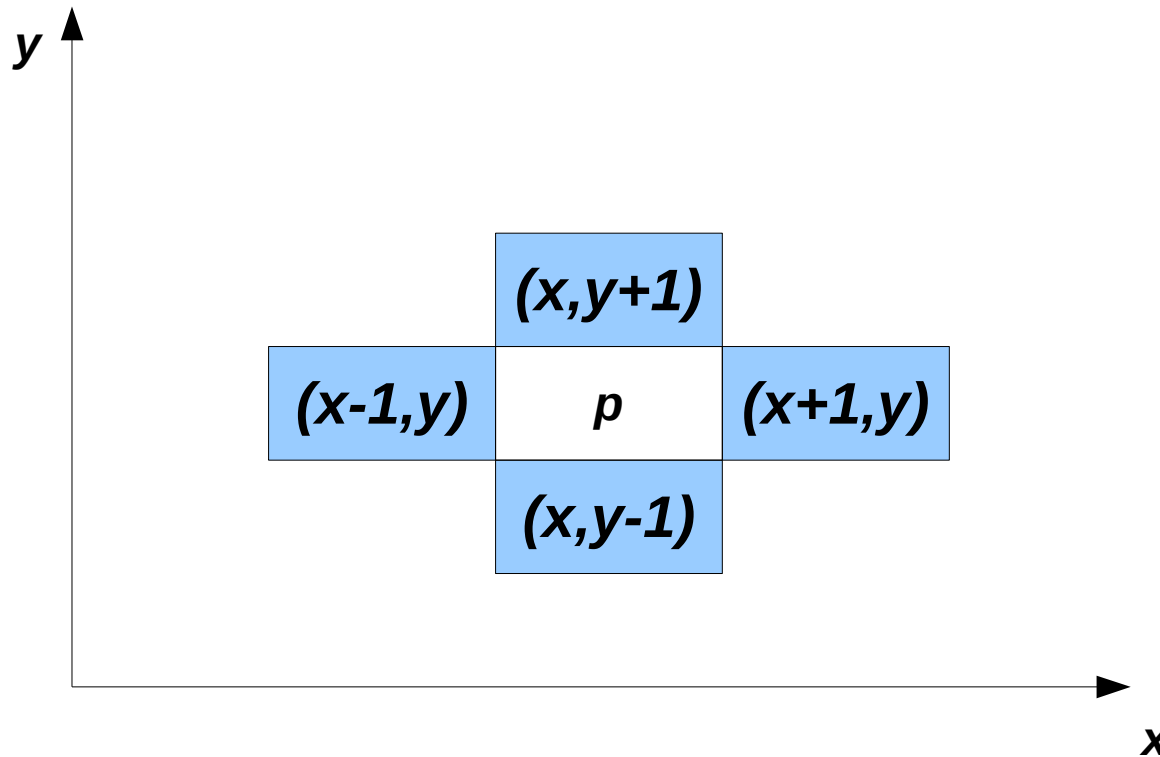
210x210

Bicubic interpolation
`imresize(a,1.5,'bicubic')`

Basic Relationships Between Pixels

- Neighbors of a Pixel
- Adjacency, Connectivity, Regions, and Boundaries
- Distance Measures
- Image Operations on a Pixel Basis
- Linear and Nonlinear Operations

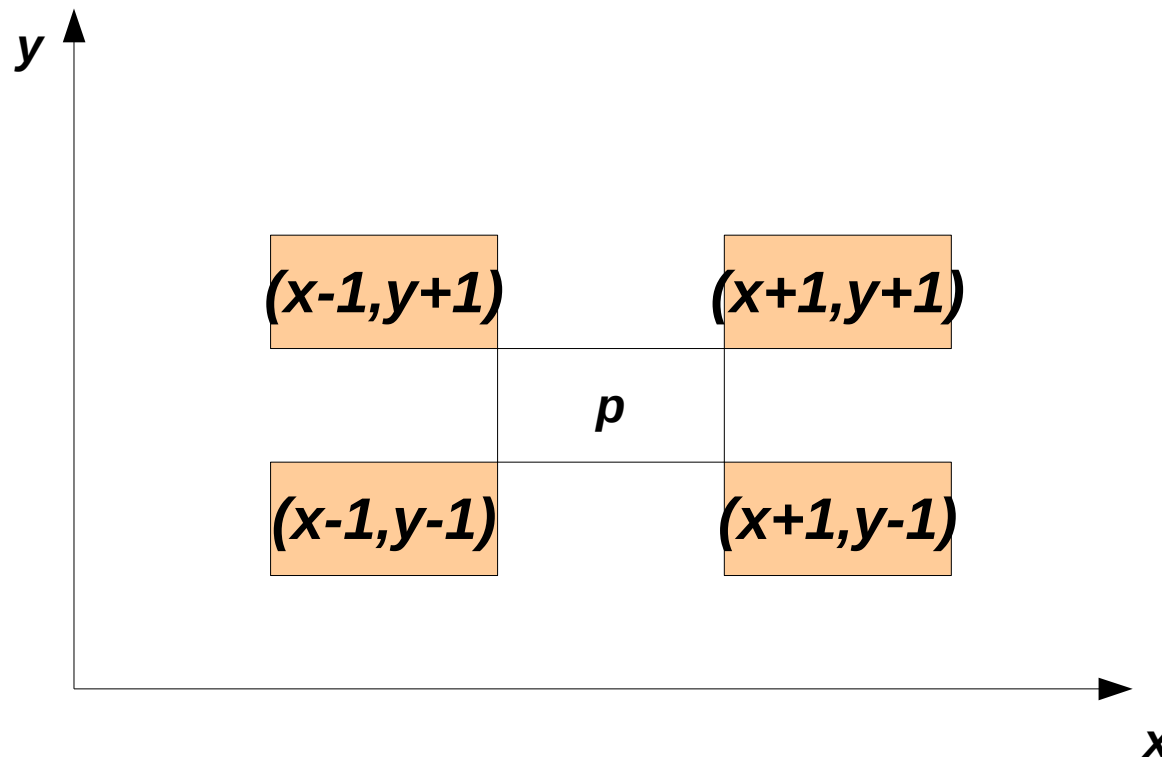
4 Neighbours of a Pixels: $N_4(p)$



A pixel p at coordinates (x, y) has four horizontal and vertical neighbors whose coordinates are given by $(x+1, y)$, $(x-1, y)$, $(x, y+1)$, $(x, y-1)$

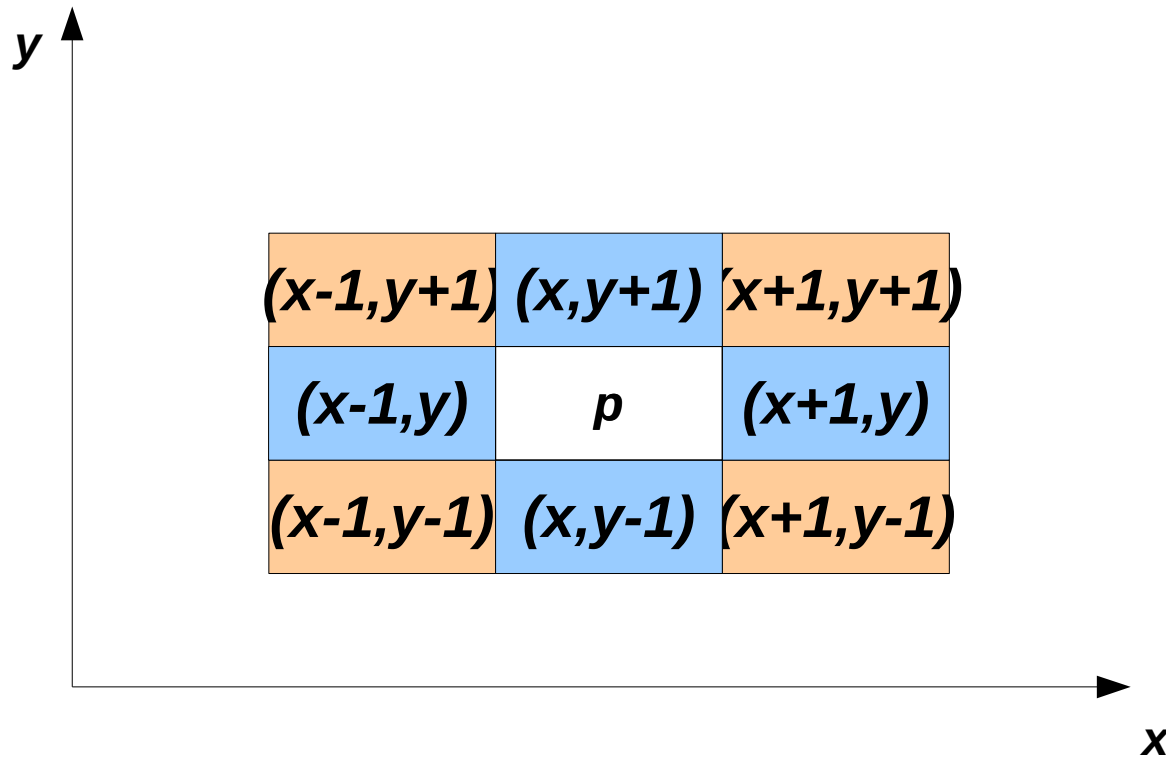
Some of the neighbors of p lie outside the digital image if (x, y) is on the border of the image.

4 Neighbours of a Pixels: Diagonal Neighbours- $N_D(p)$



The four diagonal neighbors of p have coordinates $(x+1, y+1)$, $(x+1, y-1)$, $(x-1, y+1)$, $(x-1, y-1)$

8 Neighbours of a Pixels: $N_8(p)$



Adjacency, Connectivity, Regions, and Boundaries

- Concept of Connectivity between pixels simplifies digital image concepts such as regions and boundaries
- To *establish if two pixels are connected*, it must be determined if they are *neighbors* and if their *gray levels satisfy* a specified criterion of similarity
- V will be used to denote the set of gray-level values used to define adjacency
Ex: For a binary image $V=\{0,1\}$, and so on

Types of pixel adjacency

- 1) **4-adjacency**: Two pixels p and q with values from V are 4-adjacent if q is in the set $N_4(p)$
- 2) **8-adjacency**: Two pixels p and q with values from V are 8-adjacent if q is in the set $N_8(p)$
- 3) ***m-adjacency (mixed adjacency)***: Two pixels p and q with values from V are m -adjacent if
 - (i) q is in $N_4(p)$, or
 - (ii) q is in $N_D(p)$ and the set " $N_4(p) \cap N_4(q)$ " has no pixels whose values are from V

4-neighbours $N_4(p)$



$V=\{0,1\}$

8-neighbours $N_8(p)$



$V=\{0,1\}$

m-adjacency

	2		
1	^P 1	2	
	0		

$$V=\{0,1\}$$

$$N_4(p)=\{0,1,2,2\}$$

		0	
	2	^Q 1	1
		2	

$$V=\{0,1\}$$

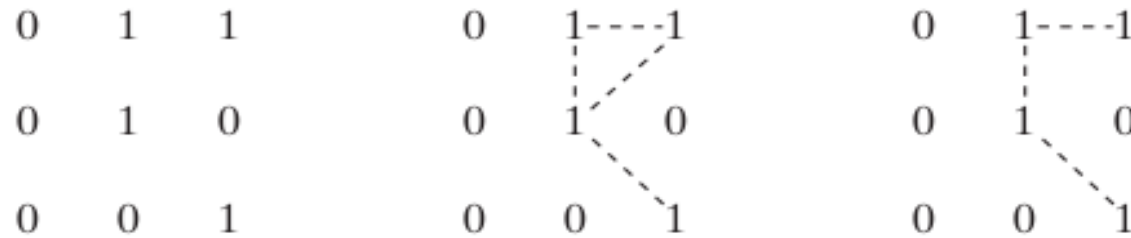
$$N_4(q)=\{2,2,0,1\}$$

		0	
	2	^Q 1	1
1	^P 1	2	
	0		

$$V=\{0,1\}$$

$$N_4(p) \cap N_4(q) = \{2,2\} \not\subseteq V$$

Types of pixel adjacency



(a) Arrangement of pixels; (b) pixels that are 8-adjacent (shown dashed) to the center pixel; (c) m-adjacency

- **Mixed adjacency is a modification of 8-adjacency.**
- **It is introduced to eliminate the ambiguities that often arise when 8-adjacency is used**

Image Subset adjacency

- **Two image subsets S1 and S2 are adjacent if some pixel in S1 is adjacent to some pixel in S2.**

Digital path and pixel connectedness

Digital Path: A (digital) path (or curve) from pixel p with coordinates (x, y) to pixel q with coordinates (s, t) is a sequence of distinct pixels with coordinates

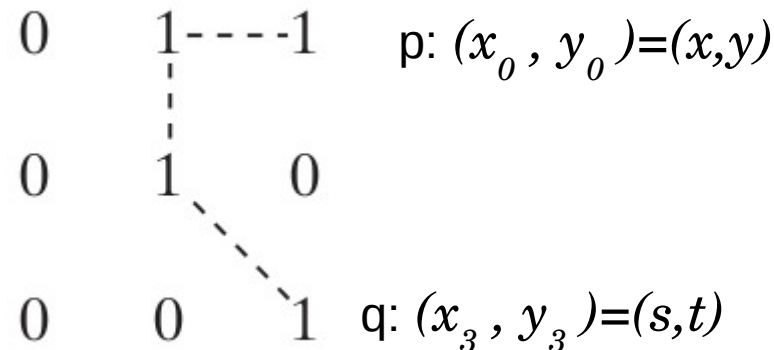
$$(x_0, y_0), (x_1, y_1), \dots, (x_n, y_n)$$

Where $(x_0, y_0) = (x, y)$; $(x_n, y_n) = (s, t)$

And pixels (x_i, y_i) and (x_{i-1}, y_{i-1}) are adjacent for $1 \leq i \leq n$

“n” is the length of the path

If $(x_0, y_0) = (x_n, y_n)$, then the path is a ***closed path***



Connected Set “S”

Let **S** represent a subset of pixels in an image.

Two pixels p and q are said to be connected in S if there exists a *path between them* consisting *entirely* of pixels in S .

For any pixel p in **S**, the set of pixels that are connected to it in **S** is called a *connected component* of **S**.

If it only has one connected component, then set S is called a ***connected set***.

Boundary of a region R

Let R be a subset of pixels in an image.

We call R a *region of the image*, if R is a **connected set**.

The boundary (also called border or contour) of a region R is the set of pixels in the region that have one or more neighbors that are not in R .

If R happens to be an entire image (which we recall is a rectangular set of pixels), then its boundary is defined as the set of pixels in the first and last rows and columns of the image.

Edges and Boundaries

The **boundary** of a finite region forms a closed path and is thus a “global” concept.

Edges are formed from pixels with derivative values that exceed a preset threshold. The idea of an edge is a “local” concept that is based on a measure of gray-level discontinuity at a point

Edges are intensity discontinuities and boundaries are closed paths.

Distance Measures

For pixels p , q , and z , with coordinates (x, y) , (s, t) , and (v, w) , respectively, D is a distance function or metric if

$$(a) D(p, q) \geq 0 \quad (D(p, q)=0 \text{ iff } p=q)$$

$$(b) D(p, q)=D(q, p) \quad \text{and}$$

$$(c) D(p, z) \leq D(p, q)+D(q, z)$$

The Euclidean distance between p and q is defined as

$$D_e(p, q) = [(x - s)^2 + (y - t)^2]^{\frac{1}{2}}.$$

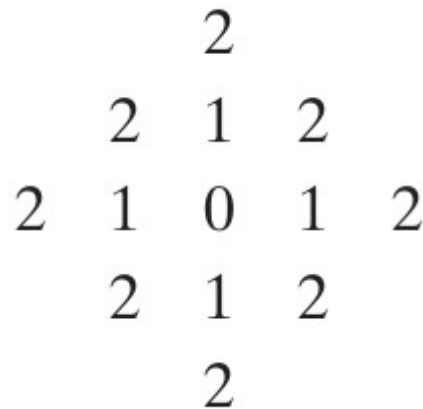
For this distance measure, the pixels having a distance less than or equal to some value r from (x, y) are the points contained in a disk of radius r centered at (x, y)

Distance Measures

The *D4 distance* (also called *city-block distance*) between p and q is defined as

$$D_4(p, q) = |x - s| + |y - t|$$

In this case, the pixels having a D_4 distance from (x, y) less than or equal to some value r form a *diamond* centered at (x, y) . For example, the pixels with D_4 distance ≤ 2 from (x, y) (the center point) form the following contours of constant distance:



The pixels with $D_4=1$ are the 4-neighbors of (x, y)

Distance Measures

The D_8 distance between p and q is defined as

$$D_8(p, q) = \max(|x - s| + |y - t|)$$

In this case, the pixels with D_8 distance from (x, y) less than or equal to some value r form a square centered at (x, y)

2	2	2	2	2
2	1	1	1	2
2	1	0	1	2
2	1	1	1	2
2	2	2	2	2

The pixels with $D_8=1$ are the 8-neighbors of (x, y)

Note: The D_4 and D_8 distances between p and q are independent of any paths that might exist between the points because these distances involve only the coordinates of the points

Distance Measures

The D_m distance between two points is defined as the shortest m-path between the points.

$$\begin{array}{cc} & p_3 \quad p_4 \\ p_1 & p_2 \\ p & \end{array}$$

Image Operations on a Pixel Basis

**Arithmetic operations on Pixel Matrix :Add/Sub/Logical
Division?**

Linear and Nonlinear Operations

Let H be an operator whose input and output are images.

H is said to be a ***linear operator*** if, for any two images f and g and any two scalars a and b ,

$$H(af + bg) = aH(f) + bH(g).$$

An operator that fails the test of above Eq is by definition ***nonlinear***.